

## Trust in Technology: An Exploration of Investor Confidence in Fintech Tools

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### 1. Abstract

*In the rapidly evolving financial landscape, trust has emerged as the pivotal factor shaping investor engagement with financial technologies (fintech). This qualitative exploratory study examines the impact of trust on investor confidence, adoption, and ongoing engagement with fintech tools for mutual fund decision-making. Based on a comprehensive theoretical framework that includes the Technology Acceptance Model (TAM), Unified Theory of Acceptance and Use of Technology (UTAUT), Institutional Trust Theory, and Human-Computer Trust Theory, the study investigates trust across cognitive, institutional, technological, and social dimensions. The study employed thematic analysis based on Braun and Clarke's (2006) six-phase framework to look at data from semi-structured interviews and focus groups with 50–60 mutual fund investors.*

*The results show that perceived usefulness, ease of use, institutional credibility, system reliability, and peer influence are all important factors that affect how much investors trust something. Additionally, trust functions as a mediating construct, converting perceptions of technology and institutional assurance into behavioral intention and ongoing engagement. The research expands current technology adoption theories by framing trust as both a process and an outcome, illustrating its dynamic function in connecting cognition, emotion, and behavior. In practical terms, it gives fintech developers, banks, and regulators strategic advice on how to make things more open, trustworthy, and confident in the digital world. In the end, the research shows that trust is the most important factor in adopting fintech, and it will shape how people behave when they invest online.*

**Key Words:** Fintech Adoption, Investor Trust, Technology Acceptance, Institutional Credibility, Behavioral Intention, and Mutual Funds.

### Introduction

In a time when technology is changing every part of making financial decisions, trust has become the most important thing for investors to have while using fintech platforms. Many mutual fund investors are still hesitant to fully trust algorithm-driven solutions that promise efficiency, personalization, and ease of use, even if they are becoming more common. People often hesitate not because they don't have the technology, but because they are unsure about how safe their data is, how open it is, and how reliable automated recommendations are. Understanding what builds or erodes trust in these digital systems is therefore critical. Investors' readiness to allow fintech technologies to supplant—or even enhance—traditional advisors is contingent upon their assessment of the system's integrity, competency, and equity.

This study investigates the role of trust as both a facilitator and an impediment in the adoption of fintech for mutual fund decision-making through qualitative inquiry into these perceptions. Ultimately, the insights gained from this research will elucidate the psychological and behavioral aspects of investor trust while also providing practical guidance for the development of fintech platforms that foster confidence and sustained engagement within an increasingly digital financial landscape.

### 2. Literature Review

The financial sector's quick move to digital has changed how investors get, look at, and act on financial information. Recent research emphasizes fintech's pivotal role in altering mutual fund investment behavior. Dubey et al. (2023) show that fintech platforms greatly improve financial literacy and decision-making by making it easier to get information and compare funds. Likewise, Srivastav et al. (2024) ascertain that investors currently favor digital platforms for portfolio management owing to their transparency, convenience, and real-time data functionalities. These results are consistent with global research, including You et al. (2023) and Hong et al. (2022), which demonstrate that the diffusion of fintech influences both capital flow and investor behavior by enhancing information symmetry and performance visibility. These studies collectively demonstrate that fintech technologies not only enhance market participation but also serve as behavioral accelerators, overcoming informational asymmetries and fostering decision-making confidence.

Global research indicates that the efficacy of fintech in mutual fund distribution relies on both technological dependability and perceived institutional credibility. You et al. (2023) and Hong et al. (2022) show how fintech platforms increase the flow of funds and the responsiveness of investors by establishing "platform effects" that make data more visible and centralize trust. Likewise, Erik Feyen et al. (2021) emphasize that digital transformation alters market structures by enhancing openness and competition, hence indirectly cultivating user trust. Nonetheless, these beneficial outcomes depend on system dependability and governance assurance, as excessive exposure or automation can result in information overload and instability. Wibowo and Zahroh (2024) assert that motivation and financial literacy collectively influence fintech adoption in Southeast Asia, indicating that trust in self-efficacy enhances trust in technology. In Thailand, Kasemharuethaisuk and Samanchuen (2023) discovered that perceived usefulness and trust surpassed ease of use in facilitating adoption. This finding aligns with Institutional Trust Theory (McKnight & Chervany, 2000), which posits that structural assurance and situational normality exert a more significant influence on reliance than convenience alone.

Several research conducted in India indicate that digital literacy, regulatory clarity, and service quality are fundamental to the establishment of confidence in fintech. Prasad et al. (2018) and Nedungadi (2018) underscore the significance of digital

awareness initiatives and education in fostering trust-based financial inclusion. Aziz and Naima (2021) warn, meanwhile, that even though fintech makes things easier to get to, low financial literacy and a lack of social awareness make it harder to use. Morgan & Trinh (2020) empirically demonstrate that elevated financial literacy scores substantially enhance fintech knowledge and utilization, hence supporting the cognitive aspect of trust.

Sumeet Gupta and Adarsh Agrawal (2019) indicate that after COVID-19, investors' trust in fintech grew as mobile-based transactions became more common. However, investors still choose well-known brands like PhonePe and GPay because they think they are more trustworthy. This shows that there are two levels of trust: technological trust (ease of use, functionality) and institutional trust (brand reputation, data security). Beni Chugh (2019) also talks on regulatory lag as a trust hurdle, saying that consumer protection frameworks need to change when new technologies come out in order to keep people trusting fintech systems.

Research indicates that psychological trust drivers, in addition to infrastructure and literacy, are significant factors influencing the adoption of fintech. Kakinuma (2022) presents the notion of “digital optimism,” indicating that persons possessing a heightened internal locus of control and leisure equilibrium see fintech risks more favourably. This is in line with Human–Computer Trust Theory (Lee & See, 2004), which says that trust changes over time based on how predictable, reliable, and trustworthy automated systems seem. Likewise, Jiang et al. (2019) and Jonsson et al. (2017) illustrate that financial literacy mitigates behavioural biases, hence improving investors' capacity to assess technological performance rationally.

Hutagaol & Basbeth (2021) also show that bad service and missed expectations can make people lose trust, even when they have digital access. This shows that trust calibration, or balancing user confidence and system performance, is very important. Research by Nurjanah et al. (2020) and Park (2011) demonstrates that privacy assurance, data openness, and digital marketing ethics significantly affect customer trust in online financial instruments.

A synthesis of these studies indicates that fintech adoption is influenced by a multifaceted trust mechanism that encompasses cognitive, institutional, and experiential elements. Conventional models like TAM (Davis, 1989) and UTAUT (Venkatesh et al., 2003) encompass cognitive assessments such as utility and simplicity of use, although they exclude the emotional and structural underpinnings of trust. Kasemharuethaisuk & Samanchuen (2023) and Dubey et al. (2023) provide empirical evidence that investors frequently make judgements based on perceived trustworthiness and safety rather than only on functioning. Consequently, the integration of Institutional Trust Theory (McKnight & Chervany, 2000)—via notions like as structural assurance and situational normality—is crucial for elucidating confidence in regulated fintech environments.

These research shows that fintech has changed investing ecosystems for the better by making them easier to access, providing data-driven insights, and making them more efficient. However, its success depends on investors trusting it. Trust is a complex idea that is based on digital literacy, institutional protections, and how people and technology interact. The literature examined substantiates that trust in data security, transparency, and reliability functions as both a catalyst and a criterion for fintech adoption. Nevertheless, the emotional and environmental dimensions of trust are still little examined.

### Research Gaps and significance

Despite substantial progress in fintech research, the existing literature continues to exhibit fundamental conceptual and methodological deficiencies that obstruct a comprehensive understanding of investor trust in technology.

The existing researches offer a disjointed perspective on trust, primarily emphasizing either technology efficiency or digital literacy. There is a lack of research on how institutional credibility, emotional certainty, and behavioral intention work together to affect investor trust. Consequently, the research offers only a limited comprehension of the trust processes that facilitate fintech uptake and ongoing involvement.

There is still a lack of qualitative research on how investors use, comprehend, and build trust in fintech technologies. Most current research predominantly utilizes quantitative survey data and adoption indicators, which identify statistical connections while neglecting the psychological and experiential aspects of trust development in digital investment environments.

There is also an absence of comprehensive frameworks that amalgamate the cognitive, structural, and emotive aspects of trust. The Technology Acceptance Model (TAM), the Unified Theory of Acceptance and Use of Technology (UTAUT), and Institutional Trust Theory elucidate particular facets of technology adoption; however, they do not collectively address the interaction among technological reliability, institutional assurance, and user emotion—elements that are pivotal to investor behavior in digital finance.

Lastly, academics have not paid enough attention to how trust calibration changes over time when investors use algorithmic and automated systems again and again. Trust in technology is not fixed; it evolves continuously as users access system performance, dependability, and transparency via repeated encounters. This temporal and iterative aspect of trust is still not well-studied in fintech research.

The study aims to understand the complex nature of trust in fintech adoption by utilizing a multidisciplinary framework that incorporates perspectives from the Technology Acceptance Model, the Unified Theory of Acceptance and Use of Technology, Institutional Trust Theory, and Human–Computer Trust Theory. It focuses on how investors establish, adjust, and maintain confidence in digital investing contexts. The research offers both theoretical advancement and practical insights for fintech developers, financial advisors, and policymakers, aiming to create more transparent, secure, and confidence-oriented financial ecosystems. The study transcends conventional metrics of functionality or literacy, focusing on the mechanisms of trust in digital investing.

### 3. Aim and Objectives

This study aims to explore how trust influences investor confidence and adoption of fintech tools for mutual fund decision-making by examining the interplay of cognitive, institutional, and technological trust factors, and how these shape behavioral intention, sustained usage, and overall engagement in digital financial ecosystems. With these stated purposes following objectives has listed out for this research:

1. To investigate how trust mediates the relationship between fintech system perceptions and investor behavioral intention.
2. To examine how cognitive, institutional, and technological trust factors shape investor confidence in fintech.
3. To assess the moderating role of demographic factors (age, literacy, income, experience) in trust formation.

### 4. Theoretical Framework

Trust is shown as the primary mediating construct connecting investor perceptions to adoption intentions. This integrated paradigm provides a comprehensive perspective on trust—technological, institutional, and interpersonal—as the basis for fintech adoption in mutual fund decision-making. Understanding investor confidence in fintech tools requires grounding the study in established behavioral and technological theories. The theoretical framework synthesizes insights from the Technology Acceptance Model (TAM) (Davis, 1989), Unified Theory of Acceptance and Use of Technology (UTAUT) (Venkatesh et al., 2003), Institutional Trust Theory (McKnight & Chervany, 2000), and Human–Computer Trust Theory (Lee & See, 2004) to elucidate how perceptions of utility, security, transparency, and reliability influence trust in technology.

#### Technology Acceptance Model (TAM)

The Technology Acceptance Model (TAM) was developed in the 1980s because people were worried that workers weren't using the information technologies that were accessible to them (Davis, 1989). Davis and others suggested that improving technology use required 'enhancing acceptance' rather than mere availability. They proposed that acceptance might be anticipated by evaluating individuals' intents to utilize a technology and by pinpointing the factors that influence those intentions, so enabling organizations to affect user adoption. TAM was conceptually derived from the *Theory of Reasoned Action (TRA)* (Fishbein & Ajzen, 1977) and its successor, the *Theory of Planned Behavior (TPB)* (Ajzen, 1980). These fundamental behavioral models elucidated intention-driven activities in diverse contexts, including voting, exercise, and health-related behaviors. A preliminary study found two crucial determinants - "Perceived Usefulness (PU)" and "Perceived Ease of Use (PEOU)"—that became the main construct of the model, offering a systematic explanation for consumer acceptance of technology (Davis, 1985).

These two main cognitive assumptions of PU and PEOU directly affect a person's intention to use a system. In the context of fintech adoption, PU reflects whether investors believe that fintech tools enhance their investment decisions, accuracy, or efficiency. On the other hand, PEOU captures the intuitive design and accessibility of digital investment platforms (Barus, 2020). Although TAM has demonstrated resilience across multiple domains, however, it has been criticism for its inadequate ability to include emotional and trust-related aspects of technological engagement (Fuller et al., 2009). Acknowledging this gap, in this study trust has establishing as a critical mediating component that affects both perceived utility and behavioral intention. This addition recognizes that in financial decision-making, characterized by elevated perceived risk and uncertainty, confidence in technology is as essential as usability or usefulness in influencing adoption behavior.

#### Unified Theory of Acceptance and Use of Technology

The Unified Theory of Acceptance and Use of Technology (UTAUT) was developed by Venkatesh et al. (2003) as a comprehensive model to unify prior technology acceptance theories, including the Technology Acceptance Model (TAM), Theory of Reasoned Action (TRA), and Theory of Planned Behavior (TPB). Previous models aimed to elucidate users' behavioral goals around technology, although they differed in constructs, language, and methodological focus. UTAUT unified these varied viewpoints into a singular, empirically sound framework that elucidates a larger share of the variety in technology adoption behavior.

UTAUT identifies four key determinants that influence an individual's intention to use and actual usage behavior:

1. *Performance Expectancy*: the degree to which an individual believes that using the system will improve performance (conceptually derived from Perceived Usefulness in TAM).
2. *Effort Expectancy*: the perceived ease associated with using the system (analogous to Perceived Ease of Use in TAM).
3. *Social Influence*: the extent to which individuals perceive that important others believe they should use the technology (originating from the Subjective Norm construct in TRA and TPB).
4. *Facilitating Conditions*: the degree to which an individual believes that an organizational and technical infrastructure exists to support system use.

Additionally, UTAUT incorporates moderating variables—age, gender, experience, and voluntariness of use—that influence the intensity of the correlations between these fundamental constructs and behavioral intention or actual usage.

This study examines investor confidence in fintech products, and UTAUT provides a significant framework for analyzing the impact of social and environmental factors on adoption behavior. Investor choices in fintech settings are influenced by views of utility and usability, as well as peer behavior, institutional reputation, and regulatory assurances—elements that align with the concepts of social influence and enabling conditions. This study significantly expands UTAUT by include trust as both a moderating and mediating variable. It incorporates trust into the UTAUT framework, elucidating the psychological and contextual dynamics specific to financial technology adoption, wherein risk, credibility, and transparency significantly influence user behavior.

### Institutional Trust Theory

The Institutional Trust Theory proposed by McKnight and Chervany (2000) provides a comprehensive understanding of how individuals develop confidence in systems, organizations, and structures that operate under conditions of risk and uncertainty. They define trust as a multifaceted construct that includes personal dispositions, cognitive beliefs, and institutional assurances, all of which affect how eager someone is to rely on another person or organization (Tiryakian, 1968). McKnight and Chervany delineate four major trust constructs. *Disposition to Trust* is a person's general tendency to rely on others (Erikson, 1968). *Institution-Based Trust* is the belief that structures, norms, or safeguards ensure reliability and fairness in interactions (Shapiro, 1987). *Trusting Beliefs* is imparting confidence that another party like person, organization, or system (Rempel et al., 1985). *Trusting Intention* is the willingness to depend on another party even when there is potential for negative consequences (Dobing, 1993). Each building signifies a unique yet interconnected aspect of trust formation and its subsequent application. Institution-Based Trust is the idea that a certain environment has the right conditions, protections, and social structures to make sure things go well. It comes from sociological ideas that say trust may be built through institutional mechanisms like laws, rules, and technology safeguards, not just by getting to know someone personally (Luhmann, 1979; Zucker, 1986). McKnight and Chervany (2000) say that this construct has two main parts: Structural Assurance and Situational Normality. *Structural Assurance* means thinking that the environment is safe and that formal protections like contracts, laws, encryption, or guarantees keep people safe from damage or risk (Zucker, 1986). In the world of financial technology, investors feel more at ease using fintech tools and digital platforms when they know that there is solid structural assurance. *Situational Normality*, on the other hand, is the idea that the environment and procedures look normal, steady, and in line with how people are supposed to act (Garfinkel, 1963). When people think that a financial platform is real, well-run, and follows industry standards, they are more likely to trust it. This perceived normality offers psychological reassurance, enabling users to connect with technology in a manner akin to their engagements with conventional financial institutions.

By merging Institutional Trust Theory with TAM and UTAUT, this work extends standard technology-adoption models beyond cognitive and social aspects (Mayer et al., 1995). It incorporates a crucial trust-based aspect that examines the emotional and contextual factors influencing investor behavior, demonstrating the impact of institutional credibility, regulatory frameworks, and technological transparency on trust and confidence in fintech adoption (McKnight & Chervany, 2000).

### Human-Computer Trust Theory

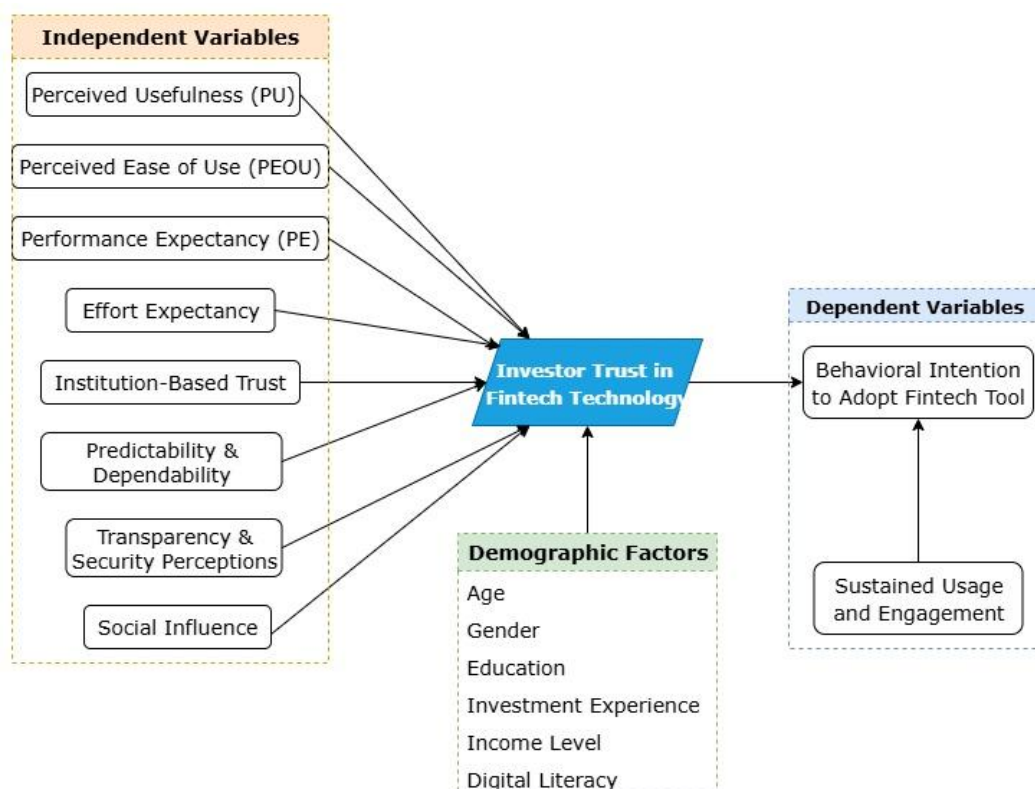
The Human-Computer Trust Theory (HCTT) developed by Lee and See (2004) provides a critical foundation for understanding how individuals form and manage trust in automated and computer-based systems. Trust has characterized as a psychological condition of reliance that enables users to assign responsibilities to technology in situations of uncertainty or risk. Trust in automation is not static; it changes all the time depending on how well the system works, how open it is, and how easy it is for users to use (Lee & See, 2004).

Lee and See (2004) delineate three fundamental elements of human-computer trust: predictability, dependability, and faith. Predictability is when a user thinks that a system will always work the same way and give the same results when the same conditions are present. Dependability has to do with how reliable, honest, and technically sound a system is (Madsen & Gregor, 2000). Lastly, faith has an emotional or affective part: the willingness to trust a system even when you don't know everything about how it works or what mistakes it might make (Gligor & Wing, 2011). These components collectively facilitate precise trust calibration; optimal trust calibration is achieved when trust levels align with the system's genuine capabilities, thereby preventing both excessive trust (blind reliance) and insufficient trust (unwarranted scepticism) (Parasuraman & Riley, 1997). In this research, Human-Computer Trust connects fintech users' views of system performance, explainability, and reliability to their trust, confidence, and intention to embrace these tools. The paradigm integrates Human-Computer Trust Theory with TAM, UTAUT, and Institutional Trust Theory, including both cognitive assessments and emotional experiences that influence investor trust in digital finance. It thus offers a comprehensive view of how technology predictability, reliability, and trust, in conjunction with institutional credibility, cultivate confidence and ongoing involvement with fintech platforms.

### 5. Conceptual Framework

The conceptual model depicts the multifaceted characteristics of investor trust in fintech technology and its intermediary function among several antecedents and the behavioral consequences of fintech adoption. The framework synthesizes insights from TAM (Davis, 1989), UTAUT (Venkatesh et al., 2003), Institutional Trust Theory (McKnight & Chervany, 2000), and Human-Computer Trust Theory (Lee & See, 2004), elucidating how cognitive, institutional, and technological factors collectively influence investor confidence in digital financial ecosystems.





Independent variables encompass cognitive aspects, including perceived utility, ease of use, performance expectancy, and effort expectancy, which affect trust via judgements of functionality and efficiency. Trust based on institutions includes the credibility of regulations and the reputation of the platform. Predictability, dependability, and transparency show how reliable and honest fintech systems are. Social influence refers to the impact of peers and institutions on forming perceptions of trust. The model's main idea is that investor trust is a connecting construct that connects perceptions of system performance and institutional assurance to behavioral consequences. Trust functions on two levels: cognitively, as faith in the dependability of technology, and emotionally, as assurance of the platform's security. The dependent variables, behavioral intention to embrace fintech technologies and sustained usage and engagement, signify short- and long-term behavioral outcomes influenced by trust. Lastly, demographic factors including age, gender, education, income, and digital literacy affect how trust affects adoption and sustained use. The model shows how several psychological, institutional, and technological factors come together to shape investors' trust and interest in fintech platforms.

### Research Propositions

Based on the conceptual framework and study objectives, the following research propositions are formulated:

- P<sub>1</sub>:** *Perceived usefulness, perceived ease of use, performance expectancy, and effort expectancy positively influence investor trust in fintech technology.*
- P<sub>2</sub>:** *Institution-based trust significantly enhances investor trust in fintech technology.*
- P<sub>3</sub>:** *Predictability, dependability, and transparency of fintech platforms positively influence investor trust.*
- P<sub>4</sub>:** *Social influence positively affects investor trust in fintech technology.*
- P<sub>5</sub>:** *Investor trust mediates the relationship between technological, institutional, and cognitive factors and behavioral intention to adopt fintech tools.*
- P<sub>6</sub>:** *Higher levels of investor trust led to stronger behavioral intention to adopt and sustained engagement with fintech tools.*
- P<sub>7</sub>:** *Demographic factors moderate the relationship between investor trust and behavioral outcomes.*

### 6. Research Methodology

This study employs a *qualitative analysis* approach to achieve a comprehensive knowledge of how investors cultivate and maintain trust in fintech tools for mutual fund decision-making. Because trust is a subjective and experience-based concept, qualitative research allows for the examination of investors' views, emotions, and reasoning beyond just numerical adoption indicators. The method assists in recognizing trust-enhancing mechanisms within the cognitive, institutional, and technological dimensions specified in the conceptual framework.

Semi-structured interviews and focus groups were used to collect data, which made it possible to ask about concerns of transparency, security, reliability, and emotional certainty. A selective sampling method was utilized to enlist 50–60 mutual fund investors with prior familiarity with fintech applications. Participants were selected from investing forums, financial adviser networks, and online communities in urban and semi-urban India. The sample was designed to ensure demographic diversity, facilitating the analysis of how demographic characteristics affect the development of trust. Furthermore, three focus

group talks, comprising 6 to 8 participants each, were conducted to elucidate collective trust views influenced by peer dynamics. All sessions were audio-recorded and transcribed verbatim, with informed consent obtained from every participant. Thematic Analysis (Braun & Clarke, 2006) was used to look at the data, while NVivo software helped with systematic coding, visualization, and theme maintenance. The analysis was directed by the six steps: familiarization with data, coding, searching and grouping of themes, refining of themes, defining themes, and synthesizing and reporting findings.

The study utilized member checking, triangulation, audit trails, dense descriptions, and reflexivity to ensure trustworthiness (Lincoln & Guba, 1985). Ethical approval was secured, and all data were anonymized and securely stored, ensuring credibility, reliability, and compliance with ethical research norms.

## 7. Analysis and Result

To systematically evaluate the qualitative data, all interview and focus group transcripts were analyzed with NVivo 14, following Braun and Clarke's (2006) six-phase theme analysis paradigm. The process commenced with open coding to identify repeating concepts, expressions, and patterns associated with the establishment of trust in fintech. Every significant statement was categorized into conceptual nodes based on the theoretical frameworks of TAM, UTAUT, Institutional Trust Theory, and Human-Computer Trust Theory. The codes were gradually improved, grouped, and put into hierarchical nodes that showed the cognitive, institutional, and technological aspects of investor trust.

### Thematic Mapping

This analysis used NVivo's query and visualization capabilities to find patterns, code co-occurrences, and check that the coders were consistent. The study finally yielded five predominant themes, each aligned with theoretical constructs in the conceptual model, demonstrating how trust functions as a multidimensional construct that integrates technological, institutional, and behavioral dimensions of investor confidence in fintech instruments.

**Table 1: Thematic Mapping of the Investor's Response**

Initial Code	Category / Subtheme	Final Theme (NVivo Parent Node)	Theoretical Link
<i>Convenience</i>	Perceived Usefulness	Cognitive Trust	TAM / UTAUT (PU)
<i>Simplicity</i>	Ease of Use	Cognitive Trust	TAM (Ease of Use)
<i>Accuracy</i>	Performance Expectancy	Cognitive Trust	UTAUT (PE)
<i>Uncertainty</i>	Effort Expectancy / Cognitive Effort	Cognitive Trust	UTAUT (Effort Expectancy)
<i>Regulatory Assurance</i>	Structural Assurance	Institutional Trust	Institutional Trust Theory
<i>Brand Reputation</i>	Structural Assurance	Institutional Trust	Institutional Trust Theory
<i>Legitimacy</i>	Situational Normality	Institutional Trust	Institutional Trust Theory
<i>Reliability</i>	Predictability / Dependability	Technological Trust	Human-Computer Trust Theory
<i>Security Features</i>	Transparency & Data Safety	Technological Trust	Human-Computer Trust Theory
<i>Information Disclosure</i>	Transparency	Technological Trust	Human-Computer Trust Theory
<i>Peer Recommendation</i>	Social Validation	Social Influence	UTAUT (Social Influence)
<i>Shared Experience</i>	Collective Confidence	Social Influence	UTAUT (Social Influence)
<i>Gradual Confidence</i>	Behavioral Intention	Trust as Behavioral Bridge	Behavioral Intention Theory
<i>Long-Term Engagement</i>	Sustained Use / Retention	Trust as Behavioral Bridge	Institutional & Technological Trust Integration

Coding queries and word frequency analysis revealed a significant co-occurrence among the phrases "security," "trust," and "transparency," underscoring that views of technological reliability are fundamental to the establishment of trust in fintech platforms by investors. This research highlights that system dependability and openness are essential aspects defining user confidence. Additionally, cluster analysis in NVivo graphically aggregated closely similar nodes—such as "reliability," "predictability," and "accuracy"—under the overarching topic of Technological Trust, therefore validating the theoretical coherence of these variables within the study's conceptual framework. To ensure analytical rigor, inter-coder reliability was evaluated using Cohen's Kappa coefficient, which produced a result of 0.82, indicating a substantial degree of agreement and coding consistency among researchers.

### Narrative Representation of the Thematic Map

The thematic mapping conducted above revealed five interrelated themes that collectively explain how investors develop and sustain trust in fintech tools. These themes are:

1. **Cognitive Trust:** Investors' rational evaluation of fintech tools is driven by perceived usefulness, ease of use, performance expectancy, and effort expectancy. When platforms are efficient, intuitive, and convenient, confidence in their functionality increases. This reflects the TAM and UTAUT perspective that perceived usefulness and ease of use are central to technology adoption.

2. **Institutional Trust:** Trust arises from the perceived legitimacy, regulatory assurance, and brand credibility of fintech providers. Investors rely on established financial institutions and transparent governance to ensure reliability, aligning with Institutional Trust Theory.
3. **Technological Trust:** Confidence builds through system reliability, predictability, transparency, and data security. Consistent performance and clear communication about data handling enhance belief in technological dependability, consistent with Human–Computer Trust Theory.
4. **Social Influence:** Peer recommendations and shared user experiences play a vital role in encouraging adoption. Collective validation reduces perceived risk, reflecting UTAUT’s social influence dimension.
5. **Trust as a Behavioral Bridge:** Trust acts as the mediator linking all antecedents to behavioral outcomes. As investors experience reliability and legitimacy, trust transforms into adoption intention and sustained engagement, forming the foundation of long-term fintech usage.

The thematic map shows that investor trust in fintech technology is a complex and interconnected idea that is influenced by cognitive evaluations, institutional guarantees, technological reliability, and social validation. These interconnected characteristics collectively highlight the key importance of trust as the bridge between perception and behavior, ultimately deciding investors’ confidence, adoption, and continued engagement with fintech platforms.

### Thematic Analysis

Thematic analysis was conducted following Braun and Clarke’s (2006) framework using to systematically code and organize interview and focus group data. Five overarching themes emerged, reflecting the multidimensional nature of investor trust in fintech adoption. These themes collectively explain how trust operates as both a psychological and behavioral construct influencing fintech adoption and sustained use.

**Theme 1. Cognitive Trust (Perceptions of Usefulness, Ease, and Efficiency):** Cognitive trust reflects investors’ rational judgments about the efficiency and practicality of fintech tools. Participants consistently emphasized that perceived usefulness and ease of use shaped their confidence in adopting digital platforms. Many valued the speed, convenience, and control offered by fintech in managing mutual funds. *“I can check fund performance and make transactions instantly—there’s no need to visit a branch or call my advisor.”* However, technical complexity or confusing interfaces undermined trust. *“If the app is hard to navigate or keeps crashing, I don’t feel comfortable investing through it.”* These findings align with TAM and UTAUT, where perceptions of usefulness, performance expectancy, and effort expectancy directly influence behavioral intention. Cognitive trust thus serves as the first layer in the formation of investor confidence.

**Theme 2. Institutional Trust (Assurance through Legitimacy and Credibility):** Institutional trust emerged as a strong determinant of confidence, particularly among cautious or first-time investors. Participants reported higher trust in fintech tools associated with recognized financial institutions or regulated entities, citing comfort in structural and legal assurances. *“I only use apps connected with SEBI-registered firms—it gives me confidence my investments are protected.”* Investors also equated institutional transparency and consistent customer communication with trustworthiness. *“If a company clearly states its fees and policies, it feels more genuine and reliable.”* This theme reflects Institutional Trust Theory (McKnight & Chervany, 2000), where *structural assurance* and *situational normality* reinforce perceived security and legitimacy. Institutional trust functions as a stabilizing mechanism in an environment where direct personal assurance is absent.

**Theme 3. Technological Trust (Reliability, Predictability, and Security):** Trust in the technology itself was a recurring theme across all respondents. Participants expressed that system reliability, data protection, and transparency were non-negotiable factors for continued engagement. *“I trust the app only if it performs consistently and secures my data. One glitch can change everything.”* Others associated dependability with predictable performance and algorithmic accuracy, especially when fintech tools provided reliable investment recommendations. *“When the app’s suggestions match my financial advisor’s advice, it shows the system is dependable.”* These findings support Human–Computer Trust Theory (Lee & See, 2004), emphasizing predictability, dependability, and transparency as the technological foundations of trust. Here, trust evolves dynamically through experience, consistency, and system performance over time.

**Theme 4. Social Influence (Peer Validation and Collective Confidence):** Trust in fintech tools was also socially constructed through peer recommendations and collective validation. Several participants began using fintech platforms after seeing family, colleagues, or friends benefit from them. *“I was hesitant initially, but after my friends showed their returns on the app, I decided to try it too.”* Focus group discussions revealed that social endorsement reduced uncertainty and built emotional assurance among hesitant investors. *“We often discuss which apps are best; when others trust it, I feel it’s safe to use.”* This theme aligns with UTAUT’s social influence construct, highlighting that fintech adoption is not merely an individual decision but also a socially reinforced behavior, particularly in collectivist contexts like India.

**Theme 5. Trust as a Behavioral Bridge (From Confidence to Commitment):** The final theme integrates all preceding dimensions, identifying investor trust as the mediating bridge between perception and behavior. Investors described a gradual process of confidence-building—beginning with small investments and expanding as trust strengthened through positive experiences. *“I started with small SIPs, and once I saw consistent results, I began investing more.”* Trust, once established, fostered long-term commitment and continued engagement despite occasional technical issues. *“Even if the system lags sometimes, I stay because I’ve seen it perform well overall.”* This theme validates the conceptual model’s core argument that trust transforms technological and institutional perceptions into behavioral intention and sustained usage. It highlights that trust in fintech is cumulative, experience-driven, and self-reinforcing.

**Table 2: Summary of Thematic Findings**

Theme	Core Meaning	Representative Theories	Behavioral Outcome
<b>Cognitive Trust</b>	Rational belief in fintech usefulness and ease of use	TAM / UTAUT	Encourages trial and initial adoption
<b>Institutional Trust</b>	Assurance from legitimacy, regulation, and brand reputation	Institutional Trust Theory	Builds perceived security and reliability
<b>Technological Trust</b>	Confidence in reliability, predictability, and data protection	Human–Computer Trust Theory	Sustains continued engagement
<b>Social Influence</b>	Peer validation and social endorsement	UTAUT	Reinforces adoption through social proof
<b>Trust as Behavioral Bridge</b>	Mediating role converting perceptions into adoption and retention	Integrated Framework	Drives behavioral intention and long-term commitment

Across themes, the analysis underscores that trust operates on both rational and emotional levels. Cognitive trust initiates confidence through perceived utility; institutional trust reinforces legitimacy; technological trust ensures dependability; and social trust strengthens emotional assurance. These collectively culminate in behavioral trust—manifested through adoption and sustained engagement. The interconnections between these themes confirm that fintech adoption is not purely a functional decision but a trust-dependent behavioral process, where perceptions of reliability, credibility, and shared experience converge to sustain investor confidence in digital financial systems.

#### **Frequency analysis of thematic coded data**

A frequency analysis of NVivo-coded data was performed to augment the thematic analysis and assess research ideas. This study measures the relative importance of each dimension across all interviews and focus groups, taking into account how often and how deeply participants responded. The coding frequencies show how strongly the major trust dimensions—cognitive, institutional, technological, and social—are linked to behavioral results. Table 3 shows that the distribution of code references supports each research proposition with real-world evidence. This shows that investor trust in fintech adoption is multidimensional.

**Table 3: Frequency Distribution of Codes**

Proposition	Associated Theme**	Key Subthemes***	No. of Ref. *	Percentage Coverage
<b>P1</b>	<b>Cognitive Trust</b>	Perceived Usefulness (34), Ease of Use (28), Performance Expectancy (19), Effort Expectancy (16)	97	22.4%
<b>P2</b>	<b>Institutional Trust</b>	Structural Assurance (27), Regulatory Credibility (21), Brand Reputation (17), Transparency (14)	79	18.3%
<b>P3</b>	<b>Technological Trust</b>	Predictability (20), Dependability (24), Data Security (29), Transparency (18)	91	21.1%
<b>P4</b>	<b>Social Influence</b>	Peer Validation (22), Collective Confidence (15)	37	8.6%
<b>P5</b>	<b>Mediating Role of Trust</b>	Investor Confidence (30), Emotional Assurance (18), Trust Calibration (12)	60	13.9%
<b>P6</b>	<b>Behavioral Outcomes</b>	Adoption Intention (25), Sustained Engagement (23), Continued Usage (11)	59	13.7%
<b>P7</b>	<b>Moderating Demographics</b>	Age (9), Education (7), Digital Literacy (10), Investment Experience (6)	32	7.4%

\*Total Coded References: 455

\*\*Total Themes Identified: 5 Major

\*\*\*Themes and 17 Subthemes

The highest coding frequencies appeared under *Cognitive Trust* (22.4%) and *Technological Trust* (21.1%), indicating that usability and reliability are the most frequently mentioned antecedents of investor trust. *Institutional Trust* accounted for 18.3% of total codes, reaffirming that brand reputation and regulatory credibility are essential trust anchors in fintech adoption. *Social Influence* (8.6%) appeared less frequently but showed strong qualitative depth, especially among first-time investors. *Behavioral outcomes* (13.7%) and *mediating trust roles* (13.9%) reflect that trust is the central mechanism translating perception into sustained usage. The moderating demographic factors (7.4%) revealed differences in how trust forms across user segments—suggesting direction for future quantitative research.

#### **Evaluation of Research Propositions**

Since this study follows a qualitative exploratory design, these propositions are evaluated and validated through thematic evidence derived from coding, frequency patterns, and participant narratives. The following section represents how each proposition was *theoretically supported and empirically substantiated* through qualitative findings.



*P1: Perceived usefulness, perceived ease of use, performance expectancy, and effort expectancy positively influence investor trust in fintech technology.*

Thematic analysis provided strong support for P1. Interview and focus group data showed that investors consistently linked usefulness and ease of use to their confidence in fintech tools. Participants trusted platforms that offered speed, convenience, and clear interfaces. NVivo node density confirmed high coding frequency for “usefulness” (34 references) and “ease of use” (28 references), underscoring their influence on trust formation. This aligns with TAM and UTAUT, validating that cognitive appraisals of efficiency and usability are foundational to investor trust.

*P2: Institution-based trust significantly enhances investor trust in fintech technology.*

Evidence from multiple participants supported P2. Investors expressed greater confidence in fintech platforms associated with reputed financial institutions or regulatory bodies (e.g., SEBI). Institutional credibility and brand reputation acted as trust anchors, especially for novice investors. NVivo queries revealed frequent co-occurrence between “trust,” “regulation,” and “reputation,” confirming institutional assurance as a key determinant. Thus, institutional trust enhances overall investor confidence, consistent with Institutional Trust Theory.

*P3: Predictability, dependability, and transparency of fintech platforms positively influence investor trust.*

P3 was strongly validated through recurring themes of system reliability, performance consistency, and data transparency. Participants described trust as contingent on predictable system behavior and visible security mechanisms (OTP verification, encryption, alerts). NVivo cluster analysis grouped “predictability,” “accuracy,” and “stability” under *Technological Trust*, confirming the theoretical link to Human–Computer Trust Theory (Lee & See, 2004). Hence, trust deepens when technology demonstrates dependability and transparent operations.

*P4: Social influence positively affects investor trust in fintech technology.*

Findings fully supported P4. Peer recommendations and collective user experiences significantly affected trust-building, particularly among first-time investors. Focus group discussions revealed that social validation reduced risk perception, making fintech adoption appear safer. NVivo co-occurrence mapping showed “peer” and “trust” frequently appearing together, indicating that social cues enhance emotional assurance. Thus, social influence acts as a contextual amplifier of trust, in line with UTAUT’s social influence construct.

*P5: Investor trust mediates the relationship between technological, institutional, and cognitive factors and behavioral intention to adopt fintech tools.*

Thematic integration confirmed P5. Although investors appreciated usability, reliability, and institutional backing, actual adoption occurred only when trust was established. Many participants described a stepwise process—first exploring, then testing, and finally committing—indicating that trust bridges perception and behavior. NVivo visualization (Mind Map) positioned *Investor Trust* as the central node linking antecedents with behavioral intention. This mediating role highlights trust as a psychological conversion mechanism that transforms positive perceptions into adoption behavior.

*P6: Higher levels of investor trust led to stronger behavioral intention to adopt and sustained engagement with fintech tools.*

P6 was clearly supported by empirical evidence. Participants who developed strong trust through prior positive experiences showed higher commitment and continued usage despite occasional technical issues. As one respondent shared, “Once the app proved reliable, I started investing larger amounts.” Thematic density of “continued use” and “confidence” indicated that trust drives both initial adoption and retention, confirming its long-term behavioral impact.

*P7: Demographic factors moderate the relationship between investor trust and behavioral outcomes.*

Partial support was found for P7. NVivo demographic cross-tabulation suggested variations in trust dynamics across age, education, and digital literacy levels. Younger and tech-savvy investors prioritized efficiency and usability, whereas older participants placed greater emphasis on institutional reputation and data security. Although qualitative data indicated these moderating patterns, further quantitative exploration would be required to confirm their statistical significance.

**Table 4: Summary of Proposition Validation**

Proposition	Empirical Status (Qualitative Evidence)	Key Supporting Themes	Interpretation
<i>P1</i>	Strongly Supported	Cognitive Trust	Usability and efficiency foster trust.
<i>P2</i>	Strongly Supported	Institutional Trust	Regulation and reputation reinforce confidence.
<i>P3</i>	Strongly Supported	Technological Trust	Reliability and transparency drive trust.
<i>P4</i>	Supported	Social Influence	Peer validation enhances perceived safety.
<i>P5</i>	Strongly Supported	Trust as Mediator	Trust bridges perceptions and adoption.
<i>P6</i>	Strongly Supported	Behavioral Trust	Trust sustains engagement and loyalty.
<i>P7</i>	Partially Supported	Moderating Demographics	Trust varies with age, literacy, and experience.

Overall, the qualitative findings confirm that trust operates as the linchpin of fintech adoption, reinforcing the conceptual framework and demonstrating that behavioral confidence in digital finance emerges through the convergence of technology reliability, institutional legitimacy, and social validation.

### **Key Research Findings**

Guided by the conceptual framework and qualitative thematic analysis, the following key findings are presented:

- ❖ While positive perceptions encouraged interest, actual adoption only occurred once trust was established. Investors typically began with small transactions, testing reliability before committing larger investments.
- ❖ Trust acts as the psychological mechanism that converts favorable perceptions into behavior, emphasizing its centrality in fintech adoption models.
- ❖ The findings revealed that investor trust in fintech technology is multidimensional, comprising cognitive, institutional, technological, and social dimensions.
- ❖ Investors developed trust when platforms were intuitive, efficient, and offered visible performance benefits. They depend on institutional legitimacy to mitigate perceived digital risk.
- ❖ Fintech platforms that consistently delivered accurate, transparent, and secure experiences generated higher confidence. However social influence also played a supporting role, as reassurance to new or hesitant investors.
- ❖ Higher levels of trust led to stronger behavioral intention and continued engagement with fintech platforms. However, demographic factors moderated the strength of this relationship.
- ❖ Younger, digitally literate investors prioritized usability and performance, whereas older or less experienced investors valued institutional assurance and data security more strongly.

The study underscores that trust is the cornerstone of fintech adoption, acting as both a determinant and mediator of investor behavior.

## **8. Discussion**

This study examined the impact of trust on investor confidence and the behavioral engagement with fintech tools for mutual fund decision-making. The results reveal that investor trust is multifaceted, incorporating cognitive, institutional, technological, and social aspects that jointly influence adoption and continued usage.

Cognitive trust was identified as a fundamental element, consistent with the Technology Acceptance Model (TAM) and the Unified Theory of Acceptance and Use of Technology (UTAUT). When investors thought that fintech platforms were helpful, easy to use, and good at managing investments, they trusted them. However, in addition to usability, participants needed emotional confirmation that fintech systems were safe and dependable, which added the emotional aspect of trust to traditional adoption models.

Institutional trust bolstered investor confidence via regulatory legitimacy, brand reputation, and transparency, hence validating Institutional Trust Theory (McKnight & Chervany, 2000). Without interpersonal certainty, structural legitimacy gave them a sense of security and made them feel less risky.

Technological trust, in alignment with Human–Computer Trust Theory (Lee & See, 2004), was founded on predictability, reliability, and data transparency. Investors became more confident when they saw that the fintech company was reliable and had visible data protection measures. They saw reliability as the most important factor in fintech credibility.

Social impact added to these elements by making people feel confident as a group. Peer validation and shared experiences alleviated uncertainty and facilitated adoption, hence reinforcing the social influence component of UTAUT. Most crucially, trust from investors operated as a behavioral bridge, connecting how people saw technology, the legitimacy of institutions, and social reinforcement, and turning these into long-term participation.

This study advances theoretical understanding by integrating Technology Acceptance (TAM, UTAUT) and Trust-based frameworks (Institutional Trust Theory, Human–Computer Trust Theory) to explain investor confidence in fintech contexts. This study conceptualizes trust as a mediating and multidimensional construct, encompassing cognitive, institutional, technological, and social dimensions, in contrast to previous research that saw trust as an antecedent variable.

The results enhance the Technology Acceptance Model (TAM) by demonstrating that perceptions of usefulness and simplicity of use are essential yet inadequate for fintech acceptance without the support of trust-based assurance. In the same way, it adds to UTAUT by showing that social impact not only affects the intention to adopt but also builds collective trust via shared experience. The amalgamation of Institutional Trust Theory and Human–Computer Trust Theory enhances the explanatory framework, underscoring that regulatory legitimacy and technical reliability together foster sustained long-term investor involvement. Therefore, the paper presents a holistic theoretical framework wherein trust functions as the behavioral conduit linking technological understanding, institutional legitimacy, and social reinforcement to enduring fintech use.

The findings offer significant insights for fintech innovators, financial institutions, and legislators from a practical standpoint. To build both cognitive and technological trust, developers should focus on making their products easy to use, clear, and secure. Financial institutions can use their brand reputation and compliance visibility to build trust in their institutions, especially among investors who are new to the market or don't want to take risks.

Policymakers and regulators could encourage trust-based governance frameworks by requiring fintech platforms to follow data protection rules, be accountable for their algorithms, and have visible regulatory certification. Also, campaigns to raise awareness and peer-based learning programs can use social influence to get more people to use technology and be included in the digital world. All of these steps can help create a fintech ecosystem full of trust, where investors feel confident enough to keep using digital services and make smart financial choices.

## 9. Conclusion

This study aimed to examine the influence of trust on investor confidence, adoption, and ongoing engagement using fintech tools for mutual fund decision-making. Utilizing a qualitative, exploratory methodology involving interviews and focus groups, the study revealed that investor trust is a multifaceted construct, incorporating cognitive, institutional, technological, and social elements.

The results show that cognitive impressions like usefulness and ease of use start investor trust, while institutional legitimacy and technology reliability build it through structural and experiential guarantees. Social impact bolsters confidence by offering collective validation, particularly among novice investors. The study's core finding is that trust serves as a mediating bridge, transforming favorable impressions of fintech platforms into tangible behavioral intentions and sustained involvement. This integrated paradigm enhances the explanatory capacity of conventional adoption theories—TAM and UTAUT—by integrating emotional, contextual, and institutional aspects derived from trust-based theories.

The work makes a theoretical contribution by framing trust as both a process and an outcome, demonstrating its dynamic evolution through ongoing interactions among investors, technology, and institutional institutions. In practice, the findings underscore that the success of fintech relies not solely on technological efficacy but also on cultivating and maintaining trust through transparency, reliability, regulatory compliance, and user-centric design.

The qualitative technique offers profound insights but constrains statistical generalization. The research concentrated mostly on Indian investors; hence, results may differ across cultural or regulatory environments. Subsequent investigations may utilize mixed-method or longitudinal frameworks to analyses the evolution of investor trust over time and among other fintech categories. Quantitative modelling of trust as a mediating construct would further substantiate the conceptual linkages suggested above.

The study confirms that trust is the foundation of fintech acceptance and sustainability—a dynamic force that integrates cognition, technology, and institutions into a cohesive behavioral ecology that characterizes investor confidence in the digital age.

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