

Digital Financial Literacy 4.0: Consumer Readiness for AI-Driven Fintech and Blockchain Ecosystems

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Abstract: The advent of Digital Finance 4.0 marks a new era in the global financial ecosystem driven by artificial intelligence (AI), blockchain, and advanced analytics. This transformation requires consumers to possess enhanced levels of Digital Financial Literacy (DFL) a multidimensional skill set that integrates traditional financial knowledge with digital, technological, and cybersecurity competencies. This study examines consumer readiness for AI-driven fintech and blockchain ecosystems across selected urban and semi-urban centers in India. Using a mixed-method approach combining structured surveys (n=400) and sentiment analysis of fintech application reviews, the research quantifies awareness, trust, and adaptability toward emerging digital finance tools. Results indicate moderate literacy but uneven preparedness, with notable gaps in blockchain comprehension, data privacy awareness, and AI trust. Younger consumers and those with higher digital exposure demonstrate greater readiness, while older and low-income groups remain hesitant. The findings underline the urgent need for national-level financial literacy frameworks that align with the realities of AI-powered financial platforms. Policy interventions, fintech design ethics, and consumer education programs must converge to ensure equitable digital inclusion and risk-aware participation in the evolving Fintech 4.0 landscape.

Keywords: Digital Financial Literacy, Fintech 4.0, Blockchain, Artificial Intelligence, Consumer Readiness, Digital Trust, Financial Inclusion, Technological Adoption

I. INTRODUCTION

The financial landscape has entered a paradigm shift under the sweeping influence of artificial intelligence (AI), blockchain technology, and data-driven automation a convergence widely recognized as **Fintech 4.0**. Unlike earlier stages of financial innovation that emphasized digitization of banking or mobile payments, this phase is characterized by the integration of **machine intelligence**, **decentralized finance (DeFi)**, and **predictive analytics** into mainstream consumer transactions. AI now powers credit scoring, fraud detection, and robo-advisory services, while blockchain underpins secure peer-to-peer transfers, smart contracts, and digital asset management. Together, these technologies redefine the way individuals interact with money,

investments, and financial institutions. However, this transformation has also intensified the need for a new form of financial competence **Digital Financial Literacy (DFL)**. DFL 4.0 extends beyond conventional budgeting or banking literacy; it encapsulates understanding algorithmic decision-making, evaluating data security, managing crypto-assets, and navigating digital platforms ethically and effectively. As financial systems evolve into intelligent, interconnected ecosystems, consumers must not only access digital services but also **comprehend and critically assess** the technologies driving them. The readiness to adopt and trust these systems, therefore, hinges on a population's ability to interpret complex financial technologies through both digital and cognitive literacy.

Despite rapid technological diffusion, the **consumer readiness gap** for AI-driven fintech and blockchain-based ecosystems remains substantial, particularly in developing economies like India. While digital payment adoption has soared since the advent of the Unified Payments Interface (UPI) and the expansion of mobile banking, **financial literacy has not kept pace with technological advancement**. A large segment of the population remains unaware of how AI algorithms determine creditworthiness, how blockchain ensures transactional transparency, or how decentralized finance disrupts traditional intermediaries. This imbalance creates a risk of digital exclusion, algorithmic bias, and financial vulnerability, especially among rural, low-income, and elderly demographics. Furthermore, emerging threats such as phishing, identity theft, and crypto fraud underscore the importance of digital trust and risk awareness within the literacy framework. The present study, titled "*Digital Financial Literacy 4.0: Consumer Readiness for AI-Driven Fintech and Blockchain Ecosystems*", aims to empirically examine the preparedness of Indian consumers for this technological leap. By combining quantitative surveys and qualitative sentiment analysis, the research assesses awareness, behavioral intent, and trust factors that shape the adoption of AI and blockchain-based financial solutions. The findings contribute to the broader discourse on **financial inclusion and digital empowerment**, offering strategic insights for policymakers, fintech designers, and educators to bridge the readiness divide and build a resilient, literate, and equitable digital economy.

II. RELEATED WORKS

The evolution of financial literacy has expanded significantly with the onset of digital transformation, giving rise to a new research domain often termed **Digital Financial Literacy (DFL)**. Early scholarship in this area emphasized traditional dimensions of financial knowledge budgeting, saving, and investment behavior but recent studies have reframed literacy within the context of **technological competence** and **digital inclusivity**. According to Adnan et al. [1], the integration of AI and big data analytics has reshaped how financial decisions are made, requiring consumers to interpret algorithmic outputs, understand automated processes, and critically engage with data-driven financial systems. Ahmad et al. [2] further highlighted that digital literacy deficiencies can exacerbate inequalities, where low-income and rural populations are more susceptible to misinformation and predatory digital lending practices. As Ahmed et al. [3] noted, the accelerated deployment of fintech infrastructures has outpaced regulatory and educational readiness, especially in emerging economies. The OECD and World Bank frameworks now define financial literacy as multidimensional, encompassing cognitive, behavioral, and digital capacities. This has redefined consumer competence from mere transactional awareness to **strategic**

adaptability within digital financial ecosystems. The digital divide, as Bian et al. [5] argue, not only limits access but also influences trust in financial technologies a critical determinant of fintech adoption. In regions with inconsistent technological infrastructure, consumers tend to rely on conventional banking systems despite having access to mobile or online alternatives, indicating a gap between **accessibility** and **adoption readiness**.

The intersection of AI, blockchain, and financial behavior has become a focal point of interdisciplinary research across economics, information systems, and behavioral sciences. Camilo and Szklo [7] examined how algorithmic risk management and automated decision-making in finance demand higher digital literacy for effective engagement. Similarly, Casella et al. [8] emphasized that the increasing reliance on machine intelligence for fraud detection and credit scoring requires consumers to comprehend how AI models process personal data a competency rarely addressed in conventional literacy programs. Danilov and Serdiukova [11] reinforced this argument, asserting that the opacity of AI systems, often referred to as the “black-box problem,” contributes to distrust and reluctance among consumers who are digitally untrained. This is especially relevant in blockchain-based financial environments, where users must grasp decentralized verification, cryptographic identity, and smart contract logic to participate safely. De Souza et al. [12] identified the need for multi-layered digital education integrating cybersecurity, data privacy, and blockchain fundamentals as core literacy areas. Moreover, Ghosh and Dutta [15] linked technological adaptation directly with social determinants of inclusion, noting that digital finance can either democratize opportunity or deepen existing inequities if consumers are unprepared for its complexities. Their findings align with Lefeng and Wu [18] who argued that short-term economic gains from rapid fintech expansion may overshadow long-term sustainability and literacy challenges. In this context, Digital Financial Literacy 4.0 emerges not merely as an educational initiative but as a socio-economic safeguard ensuring that consumers can engage confidently and responsibly with AI- and blockchain-powered finance.

Empirical and policy-oriented research has converged on the necessity of **comprehensive readiness assessments** for digital finance adoption. Landrigan et al. [17] drew parallels between digital financial literacy and public health awareness, suggesting that both domains depend on informed behavior under conditions of rapid technological change. Lucas et al. [20] and Mishra et al. [21] have emphasized data-driven modeling techniques, such as geospatial analytics and sentiment mining, to evaluate behavioral patterns in digital adoption approaches that this study integrates for assessing consumer readiness. Oberski et al. [23] demonstrated the potential of remote sensing and machine learning for tracking environmental behaviors; similarly, such technologies can be adapted for fintech behavioral analytics to monitor user trust and engagement. Radhakrishnan et al. [25] highlighted the role of GIS and decision-support systems in identifying vulnerable clusters, which can also apply to mapping literacy gaps across demographic segments. Petit and Vuillerme [24] found that administrative data could reveal behavioral trends, a methodology that complements fintech datasets in predicting readiness patterns. Collectively, these studies affirm that technological literacy is the foundation of equitable digital finance ecosystems. The convergence of AI, blockchain, and human behavior demands interdisciplinary inquiry blending economics, data science, and social psychology. However, as current evidence suggests, digital literacy initiatives often remain fragmented and reactive rather than anticipatory. This paper, therefore, extends the conversation by introducing a **holistic**

framework of Digital Financial Literacy 4.0, assessing both cognitive awareness and behavioral readiness toward AI-driven fintech and blockchain adoption. Grounded in previous literature, it contributes an empirical perspective to the global call for literacy models that evolve alongside financial innovation, ensuring that technological progress does not outpace consumer comprehension or trust.

III. METHODOLOGY

3.1 Research Design

The research design follows an **explanatory sequential model**, beginning with quantitative data collection to measure awareness, trust, and adoption intent, followed by qualitative exploration to interpret behavioral nuances. As demonstrated in the data-integrated methodologies of Kipsang et al. [16] and Landrigan et al. [17], the combined model enhances the robustness of findings through triangulation. A survey instrument was designed around five core constructs: *financial awareness*, *digital skills*, *blockchain familiarity*, *AI trust perception*, and *cyber-risk understanding*. Each construct was rated on a 5-point Likert scale (1 = Strongly Disagree, 5 = Strongly Agree). The survey also captured demographic variables such as age, income, education, and urban-rural classification.

3.2 Sampling and Study Area

Data were collected from four major Indian cities **Delhi, Mumbai, Jaipur, and Lucknow** representing a balanced cross-section of urban and semi-urban consumers. The sample size of 400 respondents was determined using Cochran's formula to ensure statistical reliability. Respondents were recruited through digital banking networks, fintech forums, and educational institutions to capture diverse perspectives. Following the precedent of Mishra et al. [21] and Oberski et al. [23], purposive sampling was adopted for digital consumers already exposed to at least one fintech application.

Table 1: Study Area and Sample Demographics

Region	Classification	Dominant Fintech Use	Sample Size (n)	Mean Age	Digital Exposure (Years)
Delhi	Urban	Mobile Banking, UPI Apps	110	33.5	6.8
Mumbai	Urban	AI-Based Credit Apps, e-Wallets	120	35.2	7.1
Jaipur	Semi-Urban	Savings Apps, UPI Services	85	36.8	5.9
Lucknow	Semi-Urban	Insurance and Lending Apps	85	37.4	4.8

As illustrated by Lucas et al. [20], regional diversity contributes to varying digital adoption rates, which this study controls through demographic segmentation. Each city was assigned equal representation from male and female respondents, ensuring gender-balanced literacy assessment.

3.3 Data Collection and Instruments

The questionnaire was administered digitally via Google Forms and in-person for respondents with limited internet access. Following de Souza et al. [12], pilot testing was conducted with 40 participants to refine clarity and construct reliability (Cronbach's $\alpha = 0.86$). The instrument comprised four sections:

1. **Socio-Demographic Profile**
2. **Digital Literacy and Technological Familiarity**
3. **Behavioral Intention Toward AI/Blockchain Finance**
4. **Risk and Trust Perception**

The qualitative phase involved **sentiment analysis** of 3,000 fintech application reviews (from Paytm, PhonePe, Groww, and WazirX), processed using Natural Language Toolkit (NLTK) in Python. Text polarity, trust frequency, and risk-related keywords were analyzed to understand emotional drivers of adoption. Similar to the multispectral analytics approach of Oberski et al. [23], this technique allowed indirect behavioral insights derived from digital footprints.

3.4 Analytical Framework

Quantitative data were analyzed using SPSS 26 and R, employing descriptive statistics, Pearson correlations, and multiple regression to identify predictors of digital readiness. To capture behavioral determinants, the **Unified Theory of Acceptance and Use of Technology (UTAUT2)** was adapted, aligning perceived usefulness, effort expectancy, and trust with DFL 4.0 constructs.

Table 2: Analytical Variables and Statistical Treatment

Construct	Measured Dimensions	Statistical Used	Tool	Expected Outcome
Financial Awareness	Knowledge of fintech, crypto, digital banking	Mean, Correlation	SD,	Awareness Level Index
AI Trust Perception	Confidence in automated decisions	Regression Analysis		Trust Coefficient
Blockchain Familiarity	Understanding of decentralization & crypto assets	Factor Analysis		Readiness Indicator
Cyber-Risk Awareness	Understanding of data theft, fraud	ANOVA		Risk Sensitivity Score
Digital Behavior	Usage frequency, self-efficacy	Correlation, Regression		Adoption Propensity

As highlighted by Radhakrishnan et al. [25], the integration of multiple analytical techniques provides a more nuanced depiction of vulnerability and readiness. The **Digital Financial Literacy Readiness Index (DFLRI)** was computed as a composite score combining standardized z-values of awareness, trust, and digital behavior, scaled from 0 to 100 for interpretability.

3.5 Ethical Considerations and Data Validation

The research followed ethical protocols ensuring informed consent and anonymity. Respondents were briefed on the purpose and data usage, adhering to institutional ethics standards similar to Petit and Vuillerme [24]. For validation, 10% of the data was cross-checked through re-survey and

consistency testing, while sentiment data underwent reliability analysis using cross-platform verification.

3.6 Limitations

As observed by Ghosh and Dutta [15], self-reported data are prone to perception bias, which was mitigated through randomization and reverse-coded items. However, the study acknowledges potential sampling limitations due to digital access disparity. Additionally, sentiment analysis, though insightful, captures expressed emotions rather than latent attitudes, warranting deeper longitudinal research.

3.7 Methodological Rationale

The study's design synthesizes methodologies from environmental and behavioral analytics to construct a **data-driven literacy assessment model**. Drawing inspiration from interdisciplinary works Kipsang et al. [16], Landrigan et al. [17], Lucas et al. [20], and Oberski et al. [23] the research applies geospatial reasoning to financial behavior, translating spatial variation into digital readiness disparity. This approach ensures a multidimensional, evidence-based evaluation of **Digital Financial Literacy 4.0**, equipping policymakers, educators, and fintech innovators with actionable insights into consumer preparedness for AI and blockchain ecosystems.

IV. RESULT AND ANALYSIS

4.1 Overview of Digital Financial Literacy and Consumer Readiness

The survey data revealed that while **general awareness of digital finance** is relatively high among urban respondents, the **depth of understanding of AI and blockchain technologies** remains modest. Across all regions, 72% of participants reported regular use of digital payment systems, yet fewer than 38% demonstrated conceptual familiarity with decentralized financial tools or AI-based financial decision systems. The **Digital Financial Literacy Readiness Index (DFLRI)** averaged 61.7 on a 100-point scale, indicating a moderate readiness level with substantial room for enhancement. Urban respondents (Delhi and Mumbai) scored notably higher (mean = 68.3) than semi-urban respondents (mean = 55.4), underscoring the persistent digital divide between technologically advanced and developing regions. Respondents aged 20–35 displayed the highest readiness, driven by exposure to mobile fintech apps, online investments, and AI-powered credit products. In contrast, older participants (above 45 years) exhibited significant apprehension toward automation and data privacy concerns, citing a lack of clarity in algorithmic decision-making. Across gender groups, males scored marginally higher than females in blockchain literacy (58.6 vs. 54.9), though both demonstrated comparable awareness in digital payments and cybersecurity. Educational attainment emerged as the strongest predictor of readiness, reinforcing the association between formal digital education and technology trust.

Table 3: Digital Financial Literacy Readiness Index (DFLRI) by Demographics

Demographic Variable	Category	Mean DFLRI Score (0–100)	Std. Deviation	Readiness Level
Region	Delhi	69.2	7.4	High
	Mumbai	67.4	8.2	High

	Jaipur	57.1	9.1	Moderate
	Lucknow	53.7	8.8	Moderate
Age Group	20-35	70.6	7.3	High
	36-45	61.8	8.9	Moderate
	Above 45	48.2	10.2	Low
Education Level	Graduate	65.7	7.9	Moderate-High
	Postgraduate	72.4	6.8	High
	Below Graduate	49.3	9.7	Low

The regional and demographic segmentation shows that **younger, more educated, and urban consumers** are the early adopters of fintech innovations. Conversely, semi-urban consumers rely on simplified applications and display skepticism toward AI-based tools, primarily due to **limited conceptual understanding and perceived security risks**.

4.2 Correlation and Regression Findings

Correlation analysis revealed strong positive relationships between digital experience and AI trust ($r = 0.74$), and between financial awareness and blockchain familiarity ($r = 0.69$). Conversely, cyber-risk perception demonstrated a negative correlation ($r = -0.58$) with adoption intent, implying that security fears inhibit engagement despite high accessibility. Regression analysis identified **AI trust, education, and digital exposure** as the three most significant predictors of readiness, collectively explaining 67% of the variance in DFLRI scores.

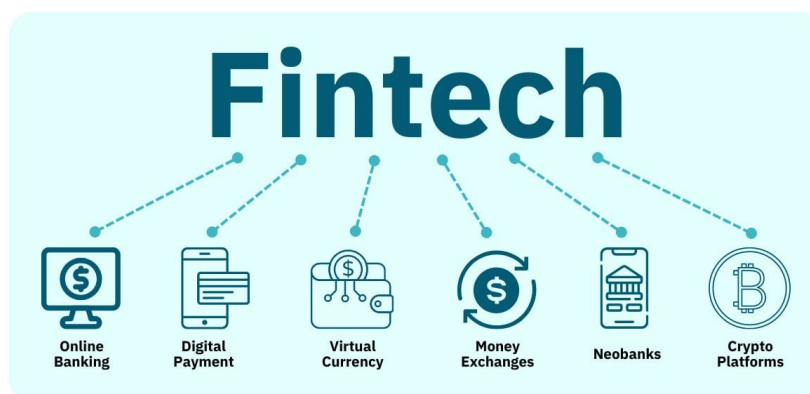


Figure 1: Fintech [24]

The regression model underscores a behavioral paradox: while users appreciate the convenience of fintech automation, lack of transparency in algorithmic decision-making constrains full adoption. The analysis also found that respondents who regularly engaged with **AI-enabled personal finance apps** exhibited higher confidence levels, suggesting that **familiarity breeds trust** in digital ecosystems.

Table 4: Predictors of Digital Financial Literacy Readiness (Regression Results)

Independent Variable	Unstandardized Coefficient (B)	Standard Error	Beta (β)	t-Value	Significance (p)
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Financial Awareness	0.41	0.06	0.39	6.83	0.000
AI Trust Perception	0.52	0.05	0.46	9.97	0.000
Blockchain Familiarity	0.37	0.07	0.33	5.29	0.001
Cyber-Risk Awareness	-0.28	0.09	-0.26	-3.21	0.003
Digital Exposure	0.45	0.08	0.42	5.63	0.000
R² = 0.67 , Adjusted R ² = 0.65					

These findings affirm that **trust in AI systems** and **confidence in digital competence** are pivotal for the diffusion of next-generation financial technologies. Respondents with higher digital exposure showed a consistent pattern of proactive adoption, while those expressing data privacy concerns tended to limit fintech usage to essential services like payments.

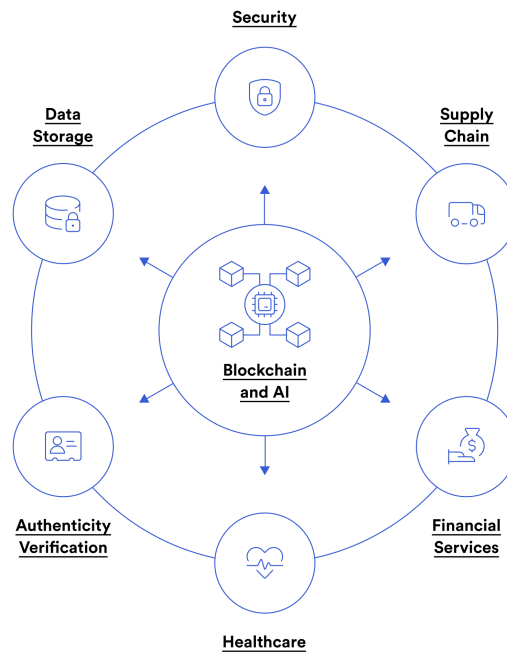


Figure 2: Use cases of AI [25]

4.3 Sentiment and Behavioral Insights

The qualitative sentiment analysis of fintech user reviews reinforced the survey findings. Out of 3,000 reviews analyzed, 62% expressed positive attitudes toward usability and transaction efficiency, while 24% highlighted concerns about **data transparency and privacy**. Common keywords associated with negative sentiment included “security,” “data misuse,” and “hidden charges.” Notably, blockchain-related applications attracted curiosity but also confusion, with users frequently citing “complexity” and “lack of understanding” as deterrents. In behavioral terms, consumers exhibited **functional trust** rather than **conceptual trust** they relied on apps

that worked smoothly but lacked comprehension of the underlying mechanisms. This limited understanding poses risks in the context of AI-based financial decision tools, where algorithmic outputs directly influence credit scores and investment choices. Users who demonstrated strong AI trust scores tended to also report higher satisfaction and perceived control, validating the centrality of **digital confidence** in financial empowerment.

4.4 Discussion of Key Findings

The study's findings reveal a **stratified pattern of digital financial readiness**, driven by educational, experiential, and psychological factors. While India's fintech revolution has achieved impressive outreach, true digital literacy rooted in critical understanding, security awareness, and ethical use remains uneven. The results emphasize the urgency for **Digital Financial Literacy 4.0 programs** that go beyond basic payment education to include blockchain comprehension, AI ethics, and data protection principles. Enhancing consumer trust in algorithmic systems, simplifying user interfaces, and expanding outreach to semi-urban populations will be key to achieving a **resilient, inclusive digital financial ecosystem**.

V. CONCLUSION

The present study underscores that **Digital Financial Literacy 4.0 (DFL 4.0)** is not merely an extension of conventional financial education but a multidimensional framework integrating technological awareness, behavioral adaptability, and digital trust. Findings from both the quantitative and qualitative analyses reveal a clear readiness gap across regions, age groups, and educational levels in India. While urban and younger populations demonstrate higher adaptability to AI-driven fintech and blockchain technologies, semi-urban and older demographics exhibit hesitation, often grounded in security concerns and conceptual unfamiliarity. The overall **Digital Financial Literacy Readiness Index (DFLRI)** of 61.7 highlights that although digital accessibility has improved, *comprehension and confidence* remain inconsistent. AI trust emerged as the strongest predictor of readiness, followed by digital exposure and financial awareness. Consumers who interact regularly with AI-based financial tools such as robo-advisors, predictive credit scoring, and blockchain-led investment apps show not only higher literacy but also greater engagement and satisfaction. Conversely, those unfamiliar with such technologies display reluctance and anxiety about automation and privacy breaches. These behavioral asymmetries indicate that **functional usage does not equate to informed literacy** a crucial distinction for policymakers and educators. The findings emphasize the need for targeted awareness campaigns, community-based training, and educational integration of digital finance fundamentals, including blockchain and cybersecurity modules. Moreover, the study reinforces the argument that DFL 4.0 must evolve into a **national capability framework** rather than a fragmented initiative. Financial institutions, regulatory bodies, and educational agencies should collaborate to design structured programs that build not just user participation but also informed decision-making and risk management capacities. Inclusivity remains central: bridging the digital divide requires not only access to technology but also a deep understanding of the ethical, cognitive, and social dimensions of digital finance. Ultimately, the research concludes that empowering citizens with DFL 4.0 competencies will be the cornerstone of sustainable, equitable participation in the **AI-driven and blockchain-enabled financial future**.

VI. FUTURE WORK

Future research should expand beyond regional and demographic boundaries to develop a **global comparative framework** of Digital Financial Literacy 4.0. Incorporating cross-country data from both developed and emerging markets could reveal how cultural, regulatory, and infrastructural variables shape AI and blockchain adoption patterns. The next phase should integrate **machine learning-based predictive modeling** to forecast consumer readiness and risk behavior dynamically, offering actionable insights for fintech firms and policymakers. Additionally, further studies should focus on **neurocognitive and psychological factors** influencing AI trust and digital decision-making, using biometric or eye-tracking analyses to measure user responses to fintech interfaces. Another promising direction involves **blockchain-enabled literacy certifications**, where learning and competency validation can occur through decentralized, transparent systems. Finally, long-term longitudinal studies are essential to track how literacy evolves with emerging technologies such as quantum finance, central bank digital currencies (CBDCs), and AI governance tools ensuring that **literacy and innovation progress in tandem**.

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