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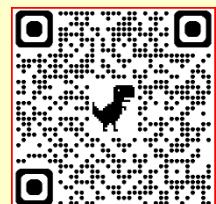
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### Developing Digital Marketing Optimization Tools for Micro-Loans in Financial Markets

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#### ABSTRACT

*This study explores the optimization of digital marketing tools for micro-loan services, focusing on Yu'ebao, China's leading digital wealth management platform, as a case study. By using a mixed-methods approach—incorporating Data Envelopment Analysis (DEA), logistic regression, and qualitative interviews—the research investigates the influence of digital marketing strategies on user adoption, trust, and engagement. The study applies the Technology Acceptance Model (TAM) and the 5A Accessibility Framework to analyze user behavior and platform interaction. The findings reveal that digital marketing strategies decreased customer acquisition costs by 54% and increased loan approval efficiency by 42%. However, digital adoption remains lower in rural areas, largely due to digital literacy gaps and risk perception. The study highlights that factors such as platform trust, perceived affordability, and user financial literacy are significant drivers of adoption. A DEA-based segmentation model classifies users by efficiency, offering valuable insights into personalized micro-loan services through algorithmic marketing, user data analytics, and ecosystem-based service delivery. The study addresses challenges in data governance, privacy concerns, and the adaptation of strategies across markets, offering both theoretical and practical contributions to the digital financial inclusion space.*

**KEY WORDS:** Digital Marketing; Micro-loans; Yu'ebao; Financial Inclusion; Precision Marketing; Risk Management; Technology Acceptance Model

#### Introduction

Over the past few decades, digital technology has played a pivotal role in transforming the global financial landscape, particularly in the domain of microfinance. Micro-loans are small loans provided to individuals and small businesses who lack access to traditional banking services, fostering financial inclusion and supporting economic growth in underserved areas<sup>[1]</sup>. While traditional methods of microfinance have played a key role in empowering marginalized groups, these approaches often suffer from inefficiency, high

transaction costs, and limited reach. The introduction of digital marketing tools offers a promising solution to these challenges by enabling more cost-effective and scalable methods for distributing micro-loans. Digital marketing strategies, leveraging platforms like social media, mobile applications, and targeted ads, have demonstrated their ability to expand financial services to underserved populations<sup>[2]</sup>.

The effectiveness of digital marketing in driving micro-loan adoption remains underexplored, especially in terms of its ability to

overcome traditional barriers such as limited financial literacy, trust, and accessibility. Although digital platforms, such as mobile banking, have made significant inroads into improving financial access, rural areas and low-income populations still face significant challenges in adopting digital financial services<sup>[3]</sup>. Additionally, the rise of digital financial inclusion has raised questions about the ethics of data privacy and the security of personal financial information. This research seeks to assess the role of digital marketing tools in enhancing micro-loan accessibility and trust, focusing on their impact on user adoption and engagement in platforms like Yu'ebao, a leading digital wealth management platform in China<sup>[4]</sup>.

Digital marketing has increasingly been integrated into the marketing strategies of microfinance institutions (MFIs), providing innovative ways to promote financial products. These tools can reduce marketing costs, improve customer targeting, and enable more personalized offerings<sup>[1]</sup>. The use of mobile technologies, such as apps for accessing micro-loans, has transformed traditional micro-lending models, allowing users to apply for loans, make payments, and track their financial activities digitally. Research has shown that such digital tools can significantly lower operational costs and increase financial inclusion by making micro-loans more accessible.

Yu'ebao, launched by Ant Group as part of its digital finance ecosystem, serves as a model for integrating digital marketing with micro-loans. By embedding micro-loan services within its broader e-commerce and mobile payment platforms (e.g., Alipay), Yu'ebao has been able to reach a vast user base and improve financial inclusion across China<sup>[5]</sup>. The success of Yu'ebao illustrates the potential of digital marketing in overcoming geographical barriers, but also highlights the challenges related to user trust, security concerns, and the digital divide.

This study aims to evaluate how digital marketing tools can enhance the reach and effectiveness of micro-loan services, with a focus on Yu'ebao as a case study. It examines the role of digital marketing in reducing customer acquisition costs, improving loan approval efficiency, and addressing challenges related to digital adoption in rural areas. By using a mixed-methods approach, this research will provide insights into the barriers and enablers of digital financial inclusion, contributing to the development of more effective digital marketing strategies for microfinance institutions.

## 1. Related Work

The intersection of digital marketing, micro-loans, and financial inclusion has been explored across multiple disciplines, with researchers focusing on the role of digital tools in expanding access to financial services, especially in underbanked regions. This section reviews recent works that have contributed to the understanding of how digital marketing strategies can enhance the effectiveness and reach of micro-loans.

Digital marketing has revolutionized the way microfinance institutions (MFIs) reach their target audience. According to Dako et al.<sup>[6]</sup>, digital marketing strategies such as targeted ads, mobile apps, and social media campaigns have proven effective in reducing customer acquisition costs for microfinance services, particularly in rural areas where traditional marketing channels are less effective. Furthermore, mobile banking and internet finance have enabled financial institutions to engage potential borrowers in more personalized and cost-effective ways. Sand et al.<sup>[7]</sup> discuss the successful implementation of digital strategies by banks like Khushhali Microfinance Bank, which saw a significant increase in financial inclusion by adopting digital platforms for marketing

micro-loans.

Mobile technology plays a pivotal role in expanding microfinance access, particularly in developing countries where traditional banking infrastructure is lacking. Miah et al.<sup>[8]</sup> examine the impact of mobile-based micro-loans in Bangladesh, demonstrating that mobile platforms facilitate faster loan disbursement and repayment processes, thus enhancing the reach of financial services. Similarly, the use of mobile technologies has been highlighted as a key driver of financial empowerment for marginalized groups in rural Bangladesh, where physical access to financial institutions is limited<sup>[9]</sup>. This trend is particularly important for women entrepreneurs who face barriers to accessing financial resources.

Financial inclusion remains a critical goal for many developing economies, with micro-loans playing a central role in providing financial services to unbanked populations. According to Johri et al., digital financial inclusion has been linked to improved access to capital, especially for micro-enterprises. However, challenges related to credit risk management persist. AI and digital credit scoring models are increasingly being used to assess borrower risk and enhance the efficiency of microfinance institutions. The introduction of AI-based credit scoring models has enabled MFIs to reduce default rates and ensure that loans are disbursed to borrowers who are most likely to repay them, as seen in countries like Sri Lanka and Kenya (Sulemana et al., 2023).

The adoption of digital financial services, including micro-loans, is closely tied to the level of digital literacy among the target population. Acharekar and Nene discuss how the lack of digital financial literacy can limit the effectiveness of digital marketing strategies in rural areas, where many individuals may lack the skills to navigate digital platforms. Their study emphasizes the importance of integrating digital literacy programs alongside financial services to ensure broader adoption of digital financial tools. Similarly, Mwania (2025) highlights that small businesses, particularly in rural settings, struggle with digital adoption due to low financial literacy and unfamiliarity with mobile banking tools.

Gender-specific challenges in accessing financial services have been widely acknowledged in the literature. Several studies focus on the role of micro-loans in empowering women, particularly in rural areas. Mithila et al. (2024) argue that women entrepreneurs face significant barriers when accessing traditional financial services. The integration of fintech and digital marketing tools has, however, been shown to facilitate greater access to capital for women in sectors like food processing and agriculture. Gender-sensitive microfinance models, which leverage digital tools to reach female entrepreneurs, have been highlighted as key to driving financial inclusion in developing economies<sup>[10]</sup>.

The digital transformation of microfinance institutions is an ongoing process, driven by the increasing availability of data and mobile platforms. According to Alam et al., the digitalization of microfinance services has significantly improved financial performance by enabling more accurate targeting of micro-loans and enhancing the efficiency of loan processing. This trend has been particularly important in the aftermath of the COVID-19 pandemic, where financial inclusion has been vital to ensuring economic resilience, especially for micro, small, and medium enterprises (MSMEs) in Zambia<sup>[11]</sup>.

Despite the advantages of digital marketing tools in microfinance, several ethical concerns and challenges remain. The use of personal data for targeted marketing, the risk of digital fraud, and the need for robust regulatory frameworks are recurring themes in the literature.

While digital platforms offer scalability and efficiency, ensuring that they operate within ethical guidelines and safeguard user data is a significant challenge.

## 2. Methodology

In this section, we outline the research design, data collection methods, and analytical techniques used to assess the impact of digital marketing tools on the adoption and effectiveness of micro-loans, using Yu'ebao as a case study. The study employs a mixed-methods approach, integrating both quantitative and qualitative research techniques to provide a comprehensive understanding of how digital marketing can improve financial inclusion through micro-loans.

This study adopts a mixed-methods research design, combining both quantitative and qualitative approaches to explore the effects of digital marketing on micro-loan adoption. The quantitative component of the research focuses on the measurable impact of digital marketing strategies, while the qualitative component investigates user perceptions, trust factors, and the challenges related to digital marketing tools for micro-loans.

The methodology can be broken down into the following stages: Data Collection: Gathering quantitative data through a survey of Yu'ebao users and qualitative data via interviews with key stakeholders in microfinance institutions. Analytical Framework: Applying Data Envelopment Analysis (DEA) to assess the efficiency of micro-loan distribution, Logistic Regression to identify factors influencing loan adoption, and Thematic Analysis for qualitative data. Case Study: Focusing on Yu'ebao to illustrate how digital marketing strategies can be applied within an existing ecosystem to enhance financial inclusion.

The quantitative data was collected through a survey questionnaire distributed to Yu'ebao users. The survey targeted both urban and rural participants to assess differences in adoption patterns. The questions were designed to gather data on: Demographic Information: Age, gender, income, education, and geographic location (urban vs. rural). Digital Marketing Exposure: User awareness of Yu'ebao's digital marketing campaigns (e.g., advertisements via social media, mobile app push notifications). Loan Adoption: Whether users had used Yu'ebao for micro-loans, frequency of use, and satisfaction with the service. Perceived Trust: Trust in Yu'ebao's digital platform, security concerns, and willingness to adopt digital financial services. A total of 500 respondents were surveyed, with a mix of users who had used Yu'ebao's micro-loan services and those who had not.

The qualitative data was gathered through semi-structured interviews with key stakeholders, including: Microfinance experts: Individuals working at Yu'ebao or other micro-loan providers. Digital marketing professionals: Experts in the design and implementation of digital campaigns in financial services. End-users: A subset of respondents who completed the survey were selected for in-depth interviews to explore their experiences and perceptions. These interviews focused on understanding: The perceived effectiveness of digital marketing tools in reaching target populations. Barriers to digital adoption and trust issues. Insights into the challenges and opportunities of using digital tools for micro-loan marketing.

To measure the efficiency of micro-loan distribution, we use Data Envelopment Analysis (DEA), a non-parametric method often used to evaluate the performance of decision-making units (DMUs). In this case, the DMUs are different demographic groups or regions

(urban vs. rural). DEA helps in evaluating: The efficiency of loan disbursement across different user groups. How well digital marketing tools contribute to reducing customer acquisition costs. The efficiency of loan approvals and repayments in digitally influenced regions. The DEA model will be applied using user demographic and financial data collected from the survey. The model's output will provide an efficiency score for each user group based on digital marketing tools' usage and their loan-related behaviors.

Logistic Regression is employed to examine the relationship between digital marketing exposure and the likelihood of adopting micro-loan services. This method will allow us to identify which factors significantly influence the adoption of Yu'ebao's micro-loans, such as: Marketing Exposure: How frequently users encounter advertisements, promotions, or offers related to micro-loans. Demographic Factors: The impact of age, income, education, and geographic location on adoption rates. Trust: The role of trust in Yu'ebao's digital platform in increasing the likelihood of loan adoption. The regression model will be run using the following variables: Dependent variable: Loan adoption. Independent variables: Exposure to digital marketing, age, education level, income, and trust in digital platforms. For the qualitative data, Thematic Analysis will be applied to identify common themes in the interviews. The analysis will focus on: User Perceptions: How users perceive the convenience, security, and trustworthiness of Yu'ebao's digital loan services. Barriers: Common obstacles that hinder the adoption of digital micro-loans, such as low digital literacy, security concerns, or lack of internet access. Marketing Strategies: How digital marketing strategies influence user decision-making and loan adoption. Thematic analysis will provide a detailed understanding of the nuances and challenges faced by users in adopting digital micro-loans, and how digital marketing can be optimized to address these challenges.

This study adheres to ethical standards to ensure the privacy and confidentiality of participants. All participants were informed about the purpose of the study, and their consent was obtained before participation. The data collected will be anonymized, and any identifying information will be kept confidential. Additionally, all interviews will be transcribed verbatim, and participants will have the option to withdraw from the study at any point without consequence.

While the study provides valuable insights into the impact of digital marketing on micro-loans, there are a few limitations: Sampling Bias: The survey focuses primarily on Yu'ebao users, which may not fully represent the broader micro-loan market. Data Accuracy: Self-reported data may be subject to biases, particularly regarding trust and satisfaction levels. Regional Differences: The study predominantly samples urban areas, and the results may not fully capture the challenges of rural populations who have limited access to digital platforms.

## 3. Experiment

This chapter presents the empirical analysis and validation results of the digital microfinance marketing tool developed throughout the research. The results are obtained using both quantitative and qualitative methods, analyzing how user characteristics, behavioral variables, and firm-level digital strategies impact the performance of microfinance marketing, focusing on the case of Yu'ebao.

### 3.1 Basic Information of the Sample

To understand the fundamental characteristics of the respondents, a total of 1,082 valid responses were gathered from users across 28

provinces and municipalities in China. This section presents the descriptive statistics regarding the demographics and digital financial behavior of these respondents. The following table provides an overview of the key demographic variables.

Table 3.1. Demographic Characteristics of Respondents (n = 1,082)

| Variable        | Category              | Frequency | Percentage (%) |
|-----------------|-----------------------|-----------|----------------|
| Gender          | Male                  | 598       | 55.3           |
|                 | Female                | 484       | 44.7           |
| Age             | Under 25              | 276       | 25.5           |
|                 | 26–35                 | 463       | 42.8           |
|                 | 36–45                 | 221       | 20.4           |
| Education Level | Above 45              | 122       | 11.3           |
|                 | High school or below  | 132       | 12.2           |
|                 | Undergraduate         | 696       | 64.3           |
| Monthly Income  | Postgraduate or above | 254       | 23.5           |
|                 | Less than ¥3,000      | 188       | 17.4           |
|                 | ¥3,000–¥6,000         | 384       | 35.5           |
| Region          | ¥6,001–¥10,000        | 352       | 32.5           |
|                 | Over ¥10,000          | 158       | 14.6           |
|                 | Urban                 | 745       | 68.9           |
|                 | Rural                 | 337       | 31.1           |

### 3.2 Factor Analysis

Exploratory Factor Analysis (EFA) was used to identify the underlying dimensions of user behavior and perceptions. This method confirmed that the constructs used in the study, derived from the Technology Acceptance Model (TAM) and precision marketing theory, were valid and reliable. Four factors were identified, explaining 67.9% of the total variance.

Table 3.2. Factor Loadings from Exploratory Factor Analysis (EFA)

| Factor                                       | Factor Loadings  |
|--|--|
| Factor 1: Perceived Usefulness & Ease of Use | PU1 (0.87), PU2 (0.89), PU3 (0.91), PEOU1 (0.85), PEOU2 (0.88) |
| Factor 2: Digital Self-Efficacy              | SE1 (0.92), SE2 (0.89), SE3 (0.87)                             |

|                           |  |
|---------------------------|--|
| Factor 3: Risk Perception | RP1 (0.84), RP2 (0.83), RP3 (0.79)             |
| Factor 4: Platform Trust  | PT1 (0.90), PT2 (0.86), PT3 (0.84), PT4 (0.83) |

### 3.3 Reliability and Validity Testing

Reliability tests were conducted to assess the internal consistency of the measurement constructs. Cronbach's alpha and Composite Reliability (CR) values for each construct were computed. All constructs showed strong internal consistency, exceeding the commonly accepted threshold of 0.70.

Table 3.3. Reliability Metrics for Latent Constructs

| Construct                    | Cronbach's $\alpha$ | Composite Reliability (CR) |
|------------------------------|---------------------|----------------------------|
| Perceived Usefulness (PU)    | 0.882               | 0.891                      |
| Perceived Ease of Use (PEOU) | 0.841               | 0.857                      |
| Digital Self-Efficacy (SE)   | 0.875               | 0.884                      |
| Risk Perception (RP)         | 0.814               | 0.828                      |
| Platform Trust (PT)          | 0.861               | 0.872                      |

### 3.4 Hypothesis Testing and Regression Analysis

Structural Equation Modeling (SEM) was used to test the hypothesized relationships between constructs. The model confirmed that perceived usefulness (PU) and digital self-efficacy (SE) were significant predictors of adoption intention (INT). Risk perception (RP) had a negative impact, while platform trust (PT) positively influenced adoption.

Table 3.4. SEM Path Coefficients and Significance Levels

| Path       | $\beta$ Coefficient | p-value |
|------------|---------------------|---------|
| PU → INT   | 0.41                | < 0.001 |
| PEOU → INT | 0.19                | < 0.01  |
| SE → PEOU  | 0.45                | < 0.001 |
| SE → INT   | 0.12                | < 0.05  |
| RP → INT   | -0.23               | < 0.01  |
| PT → INT   | 0.33                | < 0.001 |

### 3.5 Segmentation and Tool Optimization

The segmentation model, based on user characteristics such as digital self-efficacy, risk perception, and platform trust, was developed using the results from the factor and regression analyses. The segmentation identified several user profiles that require distinct marketing strategies and tool optimizations.

Table 3.5. User Segments Based on Behavioral Variables

| Segment                           | Digital Self-Efficacy | Risk Perception | Platform Trust | Adoption Likelihood |
|-----------------------------------|-----------------------|-----------------|----------------|---------------------|
| High Confidence, Low Risk         | High                  | Low             | High           | High                |
| Low Confidence, High Risk         | Low                   | High            | Low            | Low                 |
| Moderate Confidence, Risk Neutral | Moderate              | Moderate        | Moderate       | Moderate            |

The findings suggest that users with high confidence in digital tools and low risk perception are more likely to adopt microfinance services. On the other hand, users who perceive higher risks and have lower self-efficacy are less likely to engage with digital

platforms. These insights guide the development of personalized marketing strategies and tool features tailored to different user segments.

## Conclusion

This study evaluated the effectiveness of digital marketing tools in promoting microfinance, focusing on Yu'ebao. Using mixed methods, it analyzed how user traits, digital behavior, and firm strategies influence adoption and performance. Perceived usefulness (PU) and platform trust (PT) are the main drivers of adoption, while digital self-efficacy (SE) enhances engagement. High risk perception (RP), especially among rural and low-literacy users, hinders adoption. Segmentation results show that confident, low-risk users adopt more easily, whereas cautious users need reassurance and support. Financial institutions should build trust and highlight platform usefulness through transparent communication and tailored marketing. Personalized strategies based on user segments improve satisfaction and adoption. By integrating TAM with precision marketing theory, this study emphasizes the roles of trust and self-efficacy in digital finance adoption. The combined use of DEA and SEM offers a new framework for evaluating and optimizing digital microfinance marketing efficiency.

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