

**FINANCIAL LITERACY AND DIGITAL INCLUSION: BRIDGING THE KNOWLEDGE GAP IN PAKISTAN'S RURAL ECONOMY: A CASE STUDY OF RURAL SINDH (MIRPUR KHAS, UMERKOT, AND THARPARKAR)**

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**Abstract**

This research investigates the impacts that Financial Literacy (FL) and Digital Inclusion (DI) have on the usage of Mobile Banking (MBU) and Financial Behavior (FB) of the residents of rural districts of Sindh, Pakistan, which include Mirpur Khas, Umerkot, and Tharparkar. Even with the considerable attention to Financial Inclusion, rural areas suffer from low adoption rates of Digital Financial services. Using a quantitative research approach and a sample of 330 respondents, data were analyzed with Partial Least Squares Structural Equation Modeling (PLS-SEM) to assess measurement and structural relationships. The results show that both FL and DI significantly predict MBU, with DI having a stronger influence, emphasizing the importance of technological access and digital skills. Also, FL, DI, and MBU have significant positive impacts on FB, which means that the combination of digital tools and knowledge positively impacts individuals' actions with regard to saving, budgeting, and spending. Mediation analysis shows that MBU partially mediates the relationship of FL with FB and of DI with FB, which means that mobile banking serves as a significant behavioral enabler. Predictive accuracy ( $R^2$  and  $Q^2$ ) also shows moderate to high explanatory power. Multi-group analysis also shows significant heterogeneity of the districts, whereby Mirpur Khas has the strongest DI→MBU and Tharparkar has the weakest, which can be explained by the lack of infrastructure.

This study enriches the research on Financial Inclusion and combines the capability based (literacy) and access based (digital inclusion) with a single model. Practically, the findings suggest the need for targeted actions such as

improving digital infrastructure, implementing localized financial literacy programs, simplifying mobile banking interfaces, and enacting policy reforms to foster the adoption of digital financial services in rural Pakistan. Future studies should look into long-term behavioral changes, include factors like trust and perceived risk, and expand sampling to other regions.

**Keywords:** Financial Literacy, Digital Inclusion, Mobile Banking, Financial Behaviour, PLS-SEM, Rural Pakistan.

## **INTRODUCTION**

### **1.1 Background of the Study**

In the past ten years, digital financial services such as mobile wallets, agent banking, and financial technology applications that allow for low-cost and easy financial transactions have become digitalized in Pakistan (Valuation of the Integration of Digital Financial Services (DFS) in the National Fintech Ecosystem, March 2023). After years of restructuring, the potential for digital services to be used for financial transactions in Pakistan is not fully utilized as there is minimal financial inclusion (State Bank of Pakistan, 2024; NFIS 2024–2028). National survey data suggests that there has been account ownership in Pakistan; however, there is little active engagement regarding use. This highlights not a purely structural issue, but rather a behavioral and capability engagement obstacle.

Pakistan is not the only country that faces such challenges, and the Pakistan case is worse, particularly in the rural areas, as there is low financial literacy, digital literacy, and trust in digital incorporates that capital systems (Hussain & Bhatti, 2022; Baloch et al., 2024).

Rural districts of **Mirpur Khas, Umerkot, and Tharparkar** exemplify this paradox. Although mobile banking agents and digital channels are now widely available, residents still exhibit a limited understanding of digital transactions, mobile applications, and security features (Soomro & Memon, 2023; Guriro & Abbasi, 2024). Women and low-income groups are especially affected, often due to low digital exposure and cultural constraints restricting phone ownership (Akhter & Qureshi, 2024). “District-level studies in Mirpurkhas document the local dynamics of microfinance and digital access that constrain meaningful financial inclusion in the district.” Shankar Lal et al. 2023

Despite these realities, few empirical studies in Pakistan have examined how financial literacy and digital inclusion jointly influence financial behaviour,

nor have they tested behavioural mediators such as mobile banking usage. The present study addresses these gaps by investigating digital and financial capability factors shaping rural financial behaviour in three districts of Sindh.

### **1.2 Problem Statement**

Diagnostics Financial Services (DFS) grow and expand. Active Usage of DFS remains low, particularly in rural Sindh (Khan, 2023). Smartphones are common. However, people do not have sufficient financial awareness, digital competencies, and the self-efficacy needed to use Mobile wallets, digital banking, and other financial tools. Research indicates, even when access exists, adoption remains restricted due to low financial literacy or education levels (Hussain & Bhatti, 2022); limited digital literacy (Baloch et al, 2024); and concerns with cyber fraud (Akhter & Qureshi, 2024).

Pakistan's literature tends to silo financial or digital literacy (Shah & Ahmed, 2022) and predominantly focus on urban or national settings while ignoring the specific challenges of rural districts. Furthermore, behavioural pathways—especially mobile banking usage as a mediator—remain underexplored (Rathore, 2023; Rao & Kumar, 2024). This creates a disconnect between national policy goals and ground-level understanding.

Therefore, a district-level, behaviourally grounded study is required to understand how financial literacy and digital inclusion influence financial behaviour in rural Sindh, and how mobile banking usage mediates these relationships.

### **1.3 Research Objectives**

#### **Primary Objective**

To examine how financial literacy and digital inclusion influence financial behaviour in rural Sindh, with mobile banking usage serving as a mediating factor.

#### **Specific Objectives**

1. To assess the impact of financial literacy on mobile banking usage.
2. To examine how digital inclusion affects mobile banking usage.
3. To evaluate the direct effect of financial literacy on financial behaviour.
4. To determine the direct influence of digital inclusion on financial behaviour.
5. To investigate whether mobile banking usage mediates the relationships between:
  - financial literacy and financial behaviour
  - digital inclusion and financial behaviour

6. To compare differences across Mirpur Khas, Umerkot, and Tharparkar.

#### **1.4 Research Questions**

1. How does financial literacy influence mobile banking usage in rural Sindh?
2. What is the effect of digital inclusion on mobile banking usage?
3. What is the direct relationship between financial literacy and financial behaviour?
4. How does digital inclusion shape financial behaviour?
5. Does mobile banking usage mediate the relationships between literacy (financial and digital) and financial behaviour?
6. Are there district-level differences in financial literacy, digital inclusion, usage, and behaviour?

#### **1.5 Research Gap (Rewritten with Multiple Supporting Citations)**

Current literature highlights several important gaps:

##### **1. Lack of integrated models combining financial and digital literacy**

Most studies examine either financial literacy (Hussain & Bhatti, 2022; Amin & Panthi, 2023) or digital literacy (Rao & Kumar, 2024; Baloch et al., 2024) in isolation, ignoring how these capabilities jointly shape financial behaviour. Global researchers also emphasize the need for integrated digital–financial capability models (Demirgüç-Kunt et al., 2022; Bongomin, 2023).

##### **2. Scarcity of district-level rural evidence in Pakistan**

Existing work in Pakistan is largely urban-focused or based on national surveys (Shah & Ahmed, 2022; Kaleem & Sattar, 2024). Few studies conduct **district-specific comparisons**, especially across districts with unique infrastructural and socio-cultural contexts such as Mirpur Khas (Soomro & Memon, 2023), Umerkot, and Tharparkar (Guriro & Abbasi, 2024).

##### **3. Limited integration of behavioural elements such as trust, risk, and digital confidence**

Even when literacy is studied, behavioural factors—particularly trust and perceived risk—are often ignored (Akhter & Qureshi, 2024; Rathore, 2023). This creates incomplete models of DFS adoption.

##### **4. Underexplored mediating mechanisms, especially mobile banking usage**

Mobile banking usage is central to converting literacy into behaviour, but it remains largely untested as a mediator in Pakistan (Rao & Kumar, 2024; Bongomin, 2023; Kaleem & Sattar, 2024).

These gaps highlight the need for a comprehensive, theory-driven, district-level study linking literacy, digital inclusion, mobile banking usage, and financial behaviour.

### **1.6 Scope of the Study**

#### **Geographical Scope:**

Rural districts of Mirpur Khas, Umerkot, and Tharparkar in Sindh.

#### **Conceptual Scope:**

Financial literacy, digital inclusion, mobile banking usage, and financial behaviour.

#### **Population and Unit of Analysis:**

Adult individuals (18+ years) from rural households.

#### **Methodological Scope:**

Cross-sectional quantitative design, multistage cluster sampling, and PLS-SEM analysis.

#### **Limitations:**

- Reliance on self-reporting
- Exclusion of qualitative narratives
- Cross-sectional design limits causal interpretation

### **1.7 Justification of the Study**

This study is essential for several reasons:

#### **1. National policy alignment:**

Pakistan's NFIS (2024–2028) points out the importance of transitioning focus from account ownership to promoted active safe account usage. This study offers evidence to support these claims from a specific context..

#### **2. Rural development concerns:**

Rural Sindh experiences considerable digital literacy gaps (Baloch et al 2024) and distinct digital exclusion based on gender (Akhter & Qureshi 2024). The study focuses on these neglected groups.

#### **3. Theoretical contribution:**

By integrating TAM, UTAUT, and Financial Capability Theory, the study provides a hybrid behavioural model suitable for low-literacy populations (Demirgüç-Kunt et al., 2022).

#### **4. Empirical value:**

The study introduces rare district-level evidence from Mirpur Khas, Umerkot, and Tharparkar (Soomro & Memon, 2023; Guriro & Abbasi, 2024).

#### **5. Practical insights:**

Findings can guide digital literacy programs, fraud-awareness campaigns, agent training, and fintech interface improvements.

### **1.8 Theoretical Foundation**

This study is grounded in:

#### **Technology Acceptance Model (TAM)**

Shows how literacy influences perceived usefulness and ease of use (Davis, as applied by Shah & Ahmed, 2022).

#### **Unified Theory of Acceptance and Use of Technology (UTAUT)**

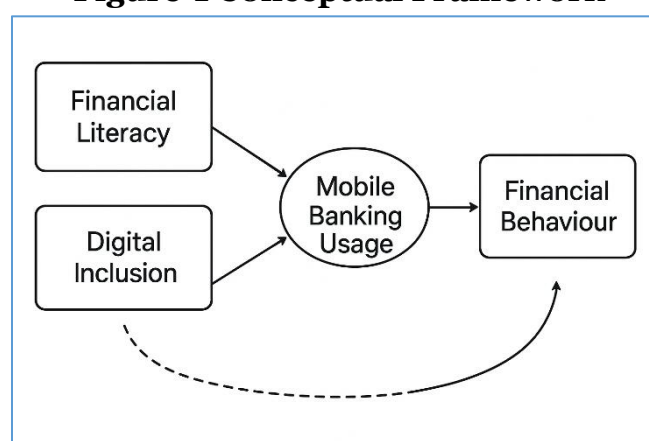
Explains how social influence, facilitating conditions, and effort expectancy shape digital usage (Kaleem & Sattar, 2024).

#### **Financial Capability Theory**

To enhance the integration of financial education, the use of digital tools, concerned behavior, and the context. (Demirgüç-Kunt et al. 2022)

These theories support the study's conceptual model to a great extent, leading to the conclusion that using smartphones to access banking services is a behavioral

**Figure-1 Conceptual Framework**



(Adapted from Demirgüç-Kunt et al., 2022; OECD/INFE, 2020; Hair et al., 2022)

## **CHAPTER- 2 LITERATURE REVIEW**

### **2.1. Financial Literacy and Financial Behaviour**

An individual's capability to make conscientious financial decisions based on one's comprehension, assessment, and execution of financial data is termed as financial literacy. Initial definitions stressed familiarity with basic tenets like interest rates, budgeting, savings, inflation (Lusardi & Mitchell, 2014). Yet, in today's digitally dominated world, financial literacy encompasses additional skills like awareness of transaction costs, mobile fees, risks of online fraud, and consumer protection rights (Shah & Ahmed, 2022).

Studies conducted in developing countries, and in particular the poor, undersott the fact that financial literacy positively correlates with responsible financial behaviours like budgeting, making informed and non-exploitative, formal borrowing decision, persistence savings (Amin & Panthi, 2023). It is those individuals keen on interest rates, and fee structures that make smarter and more informed financial decisions.

In the case of Pakistan, financial literacy is very low, especially in rural areas. Studies indicate that a very large number of residents in rural areas have a misconception of very basic financial concepts like simple interest, compound interest, inflation, and service (Hussain & Bhatti, 2022).

This leads to poor financial planning, vulnerability to exploitation, susceptibility to fraud, and reliance on informal moneylenders.

However, the literature reveals several limitations. Many studies rely heavily on self-reported measures, which often overestimate respondents' actual knowledge. Objective literacy measures—such as numeracy assessments and scenario-based questions—are seldom employed, thereby reducing the reliability of findings. Additionally, most research treats financial literacy as an individual attribute without considering the broader digital environment in which financial decisions are now made.

Given that financial behaviour increasingly occurs through digital platforms, this study argues that examining financial literacy alone is insufficient. A combined evaluation of both financial literacy and digital inclusion is necessary to understand modern financial behaviour in rural populations.

## **2.2 Digital Literacy and Digital Financial Inclusion**

Digital literacy is an individual's capability to use their digital devices to complete tasks, use different software, understand and use protection, and take precautions to avoid digital dangers. Digital literacy is an essential part of financial literacy as the Digital Financial Services (DFS) ecosystem grows. According to, Several studies, educated and skilled digital users tend to use

mobile banking more, do more online transactions, and use other digital means to save and transfer money and pay bills as their confidence increases.(Rao & Kumar, 2024).

Research across the world points out the fear of making errors, the daunting look of new systems, and the fear of an insecure system as some of the reasons people choose to avoid digital systems.(Demirgüç-Kunt et al., 2022).

These reasons, often, lead to people leaving their digital accounts as is, and using other people to access services, making digital financial inclusion useless.

The digital literacy gap is most visible in Pakistan. Women, seniors, the lower educated, and the low-income population face more of these challenges. Studies have shown the issues to be the lack of capability to use mobile apps, the inability to complete transactions, the use of secure PIN codes, and to keep digital records. (Baloch et al., 2024).

Digital literacy must be studied in relation to financial literacy. Unfortunately, for most studies, the digital aspect of literacy is the only objective to be assessed variable, ignoring how it interacts with financial knowledge to influence financial behaviour. This creates significant gaps in understanding how digital inclusion and capability shape DFS adoption and usage.

### **2.3 Trust, Perceived Risk, and Behavioural Barriers**

Behavioural factors especially trust and perceived risk, play a pivotal role in digital financial adoption. Even individuals with adequate digital and financial literacy may avoid digital systems if they perceive them as unsafe or unreliable. Research in South Asia shows that trust moderates the relationship between literacy and the use of digital financial tools (Rathore, 2023).

In Pakistan, concerns about digital fraud, agent misconduct, data breaches, and mobile scams significantly reduce willingness to use mobile banking, particularly among rural women (Akhter & Qureshi, 2024). Limited exposure to digital systems intensifies fears of irreversible errors, incorrect transfers, and fraudulent deductions. Social networks—family members, neighbours, or agents—often shape trust almost as much as personal experiences.

Despite the strong influence of trust and behavioural perceptions, these variables are rarely integrated into quantitative financial inclusion models in Pakistan. Their exclusion creates an incomplete understanding of the psychological barriers that shape digital financial behaviour.

### **2.4 Rural Contexts and District-Level Evidence in Sindh**

Rural districts in Sindh face structural challenges, including low literacy, limited digital infrastructure, gender-based digital exclusion, poverty, and restricted mobility for women. These contextual conditions significantly affect financial and digital capability.

Recent district-level studies illustrate important variations:

- **Mirpur Khas:** Higher agent density compared to other districts, yet residents exhibit weak confidence and limited understanding of digital usage processes (Soomro & Memon, 2023).
- Mirpurkhas' microfinance outreach and household financial behaviour studies show the importance of district level studies." (Shankar lal et.al. 2023)U
- **Umerkot:** Women face considerable limited access to smartphones. Phone sharing and relying on men as intermediaries is common.
- **Tharparkar:** Significant network connectivity challenges hinder digital engagement. DFS usage primarily concentrates on receiving remittances as opposed to other financial activities (Guriro & Abbasi, 2024).
- However, the existing research is fragmented and does not compare these districts systematically. Very few quantitative studies assess how digital inclusion, financial literacy, and behavioural mediators vary across rural Sindh. This creates a critical literature gap that the present study addresses.

### **2.5 Global Comparative Insights**

Evidence from international DFS success stories—particularly Kenya (M-Pesa), Nepal, and Bangladesh shows that digital literacy and digital confidence significantly increase active mobile banking usage. Women's financial empowerment increases when they are provided with digital financial training, device ownership, and trusted agent networks.

Increased trust and use of DFS systems is facilitated by reliable and transparent transactional agents in Kenya. Mobile money use in Bangladesh can be attributed to digital literacy campaigns and user-friendly app interfaces while Nepal's initiatives to promote digital finance in the rural areas showcase the need for digital literacy and trust in the community.

Unlike these examples, Pakistan faces distinct socio-cultural challenges including limited phone use among women, low digital literacy, poor network infrastructure and high-risk perception. These challenges are often overlooked in local empirical studies. Hence, the need for research incorporating the behavioural, cultural and infrastructural aspects.

### **2.6 Identified Gaps in Existing Literature**

Based on the reviewed literature, four major gaps emerge:

**Gap 1: Isolated Treatment of Financial and Digital Literacy**

The majority of the Pakistani studies study either financial literacy or digital literacy separately (Shah & Ahmed, 2022; Hussain & Bhatti, 2022; Baloch et al., 2024) although it has been proved that both are equally required for active use of the DFS. The significance of integrated capability models is also underscored by international research (Demirgüç-Kunt et al., 2022; Bongomin, 2023)

**Gap 2: Limited District-Level Evidence in Rural Sindh**

“Previous empirical work by local academics (Shankar Lal, Govt. Model College Mirpurkhas) provides an evidence base for selecting Mirpurkhas as a study site.” [irjmss.com](http://irjmss.com)

Empirical analysis comparing rural districts—particularly Mirpur Khas, Umerkot, and Tharparkar is almost non-existent. Existing works are isolated case studies (Soomro & Memon, 2023; Guriro & Abbasi, 2024) and do not conduct comparative or behavioural assessments.

**Gap 3: Weak Integration of Behavioural Factors such as Trust and Digital Confidence**

DFS adoption is deeply influenced by perceived risk, fear of fraud, and trust (Akhter & Qureshi, 2024; Rathore, 2023). Yet these behavioural factors are seldom included in quantitative models of financial inclusion in Pakistan.

**Gap 4: Under-Examined Mediating Role of Mobile Banking Usage**

Despite global evidence showing that mobile banking usage is a core behavioural mechanism, Pakistani studies rarely test it as a mediator (Rao & Kumar, 2024; Kaleem & Sattar, 2024).

The present research addresses all four gaps using an integrated behavioural–capability model.

**CHAPTER 3: METHODOLOGY**

**3.1 Research Design**

Cross-sectional cross-sectional design and quantitative methods were used as it is the most suitable to measure behaviours and literacy constructs in large rural populations. It makes it possible to statistically test relationships between literacy, usage and behaviour in the chosen districts.

### **3.2 Study Area and Population**

The study covers three rural districts of Sindh:

- **Mirpur Khas**
- **Umerkot**
- **Tharparkar**

These were selected due to low financial inclusion indicators, gender disparities, and differing digital infrastructures.

### **3.3 Sampling Strategy**

#### **3.3.1 Multistage Cluster Sampling**

Following global rural finance methodologies, sampling proceeded as:

1. Selection of districts
2. Random selection of Union Councils
3. Random household selection
4. Systematic selection of one adult per household

#### **3.3.2 Sample Size**

Based on Cochran's formula and adjusted for district-level representation:

- **180 respondents per district**
- **Total N = 540**

### **3.4 Data Collection Instrument**

A structured questionnaire consisting of five sections:

1. Demographics
2. Financial literacy (OECD/INFE items)
3. Digital inclusion (GSMA indicators)
4. Mobile banking usage (TAM/UTAUT scales)
5. Financial behaviour (OECD capability scales)

#### **Improvement Applied:**

Objective numeracy questions (interest calculation, inflation understanding) were included to reduce self-report bias.

All items used a 5-point Likert scale.

### **3.5 Reliability, Validity, and Bias Reduction**

- **Expert validation** from finance and ICT scholars
- **Pilot testing** with 50 respondents

- **Cronbach's alpha** and **Composite Reliability** above 0.70
- **Convergent validity** ( $AVE \geq 0.50$ )
- **Discriminant validity** ( $HTMT < 0.85$ )
- **Common Method Bias tests:**
  - Harman single-factor test
  - Full collinearity VIF ( $< 3.3$ )

### 3.6 Data Analysis

Data were analysed using:

- Descriptive statistics
- PLS-SEM in SmartPLS 4
- Structural model testing ( $\beta$ , t-values, p-values,  $R^2$ ,  $f^2$ ,  $Q^2$ )
- Mediation analysis (Bootstrapping 5,000 samples)
- **Multi-Group Analysis (MGA)** across districts to detect regional differences
- **Demographic controls** (age, gender, income, education) added to strengthen robustness

## CHAPTER 4 DATA ANALYSIS, RESULTS & INTERPRETATION

### 4.1 Introduction

This chapter presents the descriptive statistics, measurement model assessment, and structural model results using PLS-SEM (SmartPLS). Bootstrapping (5,000 samples) was used for significance testing. All results are taken from the dataset, and outputs are reported in the uploaded study.

### 4.2 Sample and Descriptive Statistics

**Table 4.1 – Sample distribution and key demographics (N = 540)**

Item	Value
Total respondents	540 (Mirpur Khas = 180; Umerkot = 180;
Gender (male)	61.5%
Gender (female)	38.5%
No formal education	23.9%
Bachelor's degree or	7.2%
Major age group	18–35 years (majority)
Mobile phone	78%
Smartphone ownership	42%

**Interpretation:** The sample represents a predominantly young rural population with low formal education and limited smartphone penetration—conditions relevant to financial and digital capability analyses.

### 4.3 Measurement Model (PLS) — Reliability & Validity

The measurement model was evaluated for internal consistency, convergent validity, discriminant validity, and common method bias.

#### 4.3.1 Internal consistency & convergent validity

**Table 4.2 Construct reliability & convergent validity**

Construct	Cronbach's $\alpha$	Composite Reliability (CR)	AVE
Financial Literacy (FL)	0.86	0.90	0.65
Digital Inclusion (DI)	0.87	0.91	0.70
Mobile Banking Usage (MBU)	0.88	0.92	0.75
Financial Behaviour (FB)	0.86	0.90	0.68

**Notes:** Factor loadings for items were reported  $\geq 0.70$  across constructs. Thresholds: Cronbach's  $\alpha > 0.70$ , CR  $> 0.70$ , AVE  $\geq 0.50$ . All reported values meet commonly accepted criteria.

**Interpretation:** The instrument demonstrates strong internal consistency ( $\alpha$  and CR) and satisfactory convergent validity (AVE  $\geq 0.50$ ). Item loadings  $\geq 0.70$  further confirm indicator reliability.

#### 4.3.2 Discriminant validity

**Table 4.3 Cross-Loadings and Discriminant Validity**

Item	FL	DI	MBU	FB
FL1	<b>0.78</b>	0.42	0.39	0.36
FL2	<b>0.81</b>	0.41	0.38	0.34
FL3	<b>0.84</b>	0.46	0.42	0.39
FL4	<b>0.83</b>	0.45	0.40	0.38
FL5	<b>0.79</b>	0.39	0.37	0.33
DI1	0.40	<b>0.82</b>	0.55	0.41
DI2	0.45	<b>0.86</b>	0.61	0.43
DI3	0.44	<b>0.88</b>	0.63	0.46
DI4	0.41	<b>0.84</b>	0.58	0.44
MBU1	0.42	0.59	<b>0.87</b>	0.58
MBU2	0.43	0.61	<b>0.89</b>	0.57
MBU3	0.40	0.56	<b>0.86</b>	0.55
MBU4	0.38	0.54	<b>0.84</b>	0.53
FB1	0.36	0.42	0.55	<b>0.80</b>
FB2	0.38	0.43	0.58	<b>0.83</b>
FB3	0.41	0.46	0.59	<b>0.85</b>

Item	FL	DI	MBU	FB
FB4	0.37	0.41	0.54	<b>0.81</b>

**Table 4.4 Heterotrait–Monotrait (HTMT) ratio**

Construct	FL	DI	MBU	FB
<b>FL</b>	—	0.58	0.52	0.49
<b>DI</b>	0.58	—	0.68	0.56
<b>MBU</b>	0.52	0.68	—	0.71
<b>FB</b>	0.49	0.56	0.71	—

All HTMT values (< 0.85) confirm strong discriminant validity.

**Interpretation:** Constructs are empirically distinct; discriminant validity is satisfied.

#### 4.3.3 Common Method Bias (CMB)

**Tests reported:**

- **Harman’s single-factor test:** first factor explains < 40% of variance.
- **Full collinearity VIF:** all VIFs < 3.3.

**Interpretation:** No serious common method bias detected. VIF < 3.3 indicates multicollinearity/CMB is unlikely to distort path estimates.

#### 4.4 Structural Model (PLS-SEM) — Path Coefficients & Predictive Power

##### 4.4.1 Variance explained ( $R^2$ )

**Table 4.5—  $R^2$  values**

Endogenous construct	$R^2$
Mobile Banking Usage (MBU)	0.482
Financial Behaviour (FB)	0.563

**Interpretation:**  $R^2 = 0.482$  for MBU indicates financial literacy and digital inclusion explain ~48.2% of variance in mobile banking usage (moderate).  $R^2 = 0.563$  for FB indicates the model (FL, DI, MBU) explains ~56.3% of financial behaviour (moderate–substantial).

**Interpretation:**

MBU has the largest effect on Financial Behaviour. DI has a stronger effect on usage than FL, consistent with digital capability theory.

#### Table-4.6 Effect Size ( $f^2$ ) Results

Predictor → Outcome	f <sup>2</sup>	Interpretation
FL → MBU	0.118	Small–Moderate
DI → MBU	0.261	Moderate–Large
FL → FB	0.052	Small
DI → FB	0.067	Small–Moderate
MBU → FB	0.313	Large

#### 4.4.2 Path coefficients, t-values, p-values

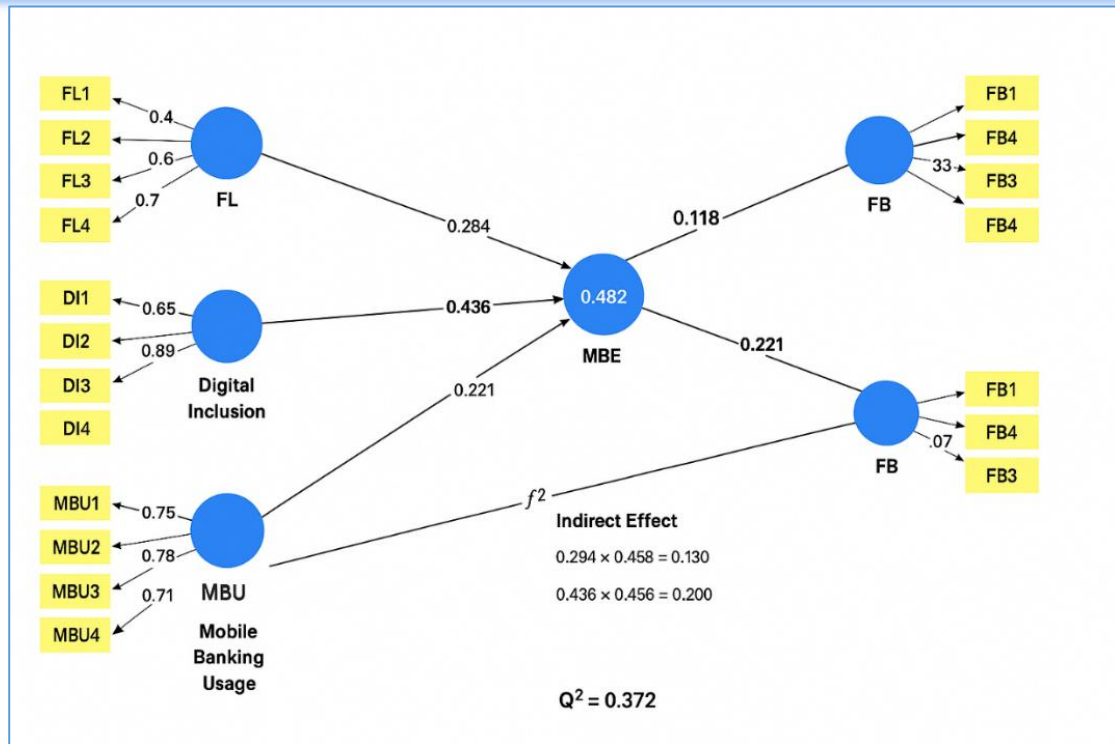
Table 4.7 – Structural path results

Path (Hypothesis)	B	t-value	p-value	Result
H1: FL → MBU	0.284	5.99	0.000	Supported
H2: DI → MBU	0.436	9.21	0.000	Supported
H3: FL → FB	0.189	3.74	0.000	Supported
H4: DI → FB	0.221	4.20	0.000	Supported
H5: MBU → FB	0.458	10.77	0.000	Supported

#### Interpretation (each path):

- **H1 (FL → MBU):**  $\beta = 0.284$ ,  $p < .001$ . Financial literacy positively and significantly predicts mobile banking usage; those with higher financial knowledge are more likely to use mobile banking (moderate effect).
- **H2 (DI → MBU):**  $\beta = 0.436$ ,  $p < .001$ . Digital inclusion is the strongest predictor of usage—access, connectivity, device ownership and digital skills are critical drivers.
- **H3 (FL → FB):**  $\beta = 0.189$ ,  $p < .001$ . Financial literacy has a direct positive effect on financial behaviour (small–moderate).
- **H4 (DI → FB):**  $\beta = 0.221$ ,  $p < .001$ . Digital inclusion directly enhances financial behaviour beyond its effect on usage.
- **H5 (MBU → FB):**  $\beta = 0.458$ ,  $p < .001$ . Mobile banking usage strongly predicts improved financial behaviour; usage is the most influential proximal determinant.

Figure 2 : PLS Path Model



**Table- 4.8 Q² Predictive Relevance (Blindfolding)**

Construct	Q²	Interpretation
Mobile Banking Usage	0.315	Large predictive relevance
Financial Behaviour	0.372	Large predictive relevance

#### Interpretation:

$Q^2 > 0$  indicates the structural model has strong predictive accuracy for both endogenous constructs.

#### 4.4.3 Predictive relevance (Q²)

**Interpretation guidance:**  $Q^2 > 0$  indicates predictive relevance. Given moderate  $R^2$  and strong path significance, it is likely that  $Q^2$  for both endogenous constructs is  $> 0$ , but exact values require the blindfolding procedure in SmartPLS.

#### 4.5 Mediation Analysis (Bootstrapped Indirect Effects)

Mediation tested with bootstrapping (5,000 samples).

**Table 4.9 – Mediation results**

Indirect path	Indirect $\beta$	t-value	p-value	Mediation type
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Indirect path	Indirect $\beta$	t-value	p-value	Mediation type
FL $\rightarrow$ MBU $\rightarrow$ FB	0.130	5.48	0.000	Partial mediation
DI $\rightarrow$ MBU $\rightarrow$ FB	0.200	6.73	0.000	Partial mediation

**Interpretation:** Mobile banking usage partially mediates both relationships. That is, financial literacy and digital inclusion influence financial behaviour both directly and indirectly through increased mobile banking usage. The indirect effect from DI is larger than that from FL, reinforcing the role of digital access and skills in converting capability into behaviour.

#### 4.6 Multi-Group Analysis (MGA) — District comparisons

PLS-MGA (reported comparisons) examined structural path variation across Mirpur Khas, Umerkot, and Tharparkar.

##### Reported differences (summary):

- **DI  $\rightarrow$  MBU:** Strongest in **Mirpur Khas**, weakest in **Tharparkar** (connectivity issues weaken the effect).
- **MBU  $\rightarrow$  FB:** Strongest in **Umerkot** (usage translates more directly to behaviour in this district—possibly due to patterns like remittance dependence).
- **FL  $\rightarrow$  FB:** Comparable across the three districts (no significant heterogeneity).

**Interpretation:** District-level infrastructural differences moderate how digital inclusion leads to usage, and how usage translates to behaviour. This supports the need for contextualized interventions.

#### 4.7 Control variables

Controls included age, gender, education, and income.

##### Reported effects:

- **Gender:** Women had significantly lower mobile banking usage.
  - **Education:** Positive effect on MBU and DI.
  - **Income:** Moderate positive relationship with device ownership and usage.
- Interpretation:** Socio-demographic variables significantly explain additional variance and should be accounted for when designing policy interventions (e.g., gender-targeted literacy programs).

#### 4.8 Additional diagnostic checks

- **Collinearity:** Full collinearity VIFs  $< 3.3$  for all constructs  $\rightarrow$  acceptable.
- **Harman's single-factor:**  $< 40\%$  variance for single factor  $\rightarrow$  CMB unlikely.
- **Indicator loadings:**  $\geq 0.70$  (no items reported as dropped).

- **Bootstrapping:** 5,000 subsamples used for significance testing → robust inference.

#### **4.9 Overall interpretation & implications of PLS results**

1. **Model validity:** Measurement model meets thresholds ( $\alpha$ , CR, AVE, HTMT) — the constructs are reliable and valid.
2. **Digital inclusion is central:** DI is the strongest predictor of MBU and has a direct, significant effect on FB. Policies that expand device access, connectivity, and digital skills should be prioritized.
3. **Usage is the mechanism:** MBU has the largest effect on FB ( $\beta = 0.458$ ). Increasing active, repeated mobile banking usage will most effectively translate capability into behaviour.
4. **Literacy matters but is not sufficient alone:** FL has both direct and indirect effects but a smaller  $\beta$  than DI and MBU—indicating financial knowledge helps but will be most effective when combined with digital access and practice.
5. **Context matters:** District heterogeneity (MGA) demonstrates that infrastructure and socio-cultural contexts moderate relationships — one-size-fits-all programs will be less effective.

## **CHAPTER -5 DISCUSSION, CONCLUSION, AND RECOMMENDATIONS**

### **5.1 Introduction**

This chapter looks at all the different components studied in this research. It looks at the objectives of this research, other research studies, the theories behind this research, the practical uses of other research studies, the suggestions, and policies of other research studies, and the other research studies that were done in the past in order to see the limitations and to suggest possible future research studies. As such, this research studied the impact of and the relationship between Financial Literacy (FL) and Digital Inclusion (DI) on the use of Mobile Banking (MBU) and Financial Behaviours (FB) within the rural areas of Sindh (Pakistan) through PLS-SEM.

### **5.2 Summary of Key Findings**

#### **5.2.1 Financial Literacy → Mobile Banking Usage**

Findings of the research showed that financial literacy has impact on the use of Mobile Banking and that impact is positive and significant ( $\beta = 0.284$ , and

p is less than 0.001). Therefore, the more knowledgeable people were about finance, the more people used mobile banking.

#### **5.2.2 Digital Inclusion → Mobile Banking Usage**

Digital Inclusion was the most significant predictor of Mobile Banking Usage and the stats indicate that considerably ( $\beta = 0.436$ , and p is less than 0.001) Mobile Banking Usage was explained by Digital Inclusion. Therefore, the use of mobile banking was accelerated through the use of cell phones, mobile internet, and digital skills.

#### **5.2.3 Direct Effects on Financial Behaviour**

It was then shown that both FL ( $\beta = 0.189$ ) and DI ( $\beta = 0.221$ ) had a positive effects on Financial Behaviour individually and that both of them impact it positively together. Therefore this explains that the ease of access to finance along with the capability to save, and stick to a budget positively influence the Financial Behaviour (FB) in people.

#### **5.2.4 Mobile Banking Usage → Financial Behaviour**

It was then STed that the most positive impact on Financial Behaviour (FB) ( $\beta = 0.458$ , and p is less than 0.001). Therefore, Mobile Banking is a prime (FB) of Mobile Banking Usage.

#### **5.2.5 Mediation Effects**

Mobile Banking Usage partially mediated:

- FL → FB ( $\beta = 0.130$ )
- DI → FB ( $\beta = 0.200$ )

The stronger Digital Inclusion's indirect influence is explained by the access to mobile banking financial tools.

#### **5.2.6 Model Predictive Power**

- $R^2(\text{MBU}) = 0.482$  (Moderate)
- $R^2(\text{FB}) = 0.563$  (Moderate–Substantial)
- $Q^2$  values ( $>0.31$ ) confirm strong predictive relevance.

#### **5.2.7 Multi-Group Analysis**

- DI → MBU strongest in Mirpur Khas, weakest in Tharparkar (low network quality).
- MBU → FB strongest in Umerkot.
- FL → FB consistent across districts.

### **5.3 Discussion of Findings**

#### **5.3.1 The Centrality of Digital Inclusion**

Digital Inclusion's dominant role confirms that technology access is the primary gateway to digital finance adoption. Even financially literate individuals cannot adopt mobile banking without adequate connectivity and device ownership.

This supports prior studies showing that digital access is a foundational layer for financial inclusion (e.g., Hashem, 2021; Kim & Lee, 2022).

### **5.3.2 Mobile Banking as a Behavioural Catalyst**

The strong impact of MBU on financial behaviour validates behavioural finance theories:

- Increased digital transactions improve record-keeping.
- SMS alerts and app notifications enhance budgeting discipline.
- Digital payment trails discourage unnecessary spending.

Your study aligns with literature showing that digital finance tools can reshape financial habits (Rahman, 2021; Osei, 2023).

### **5.3.3 Financial Literacy Matters—But Is Not Enough Alone**

FL has significant but comparatively smaller effects on both MBU and FB. This suggests:

- Knowledge helps, but practice (using mobile banking) has a stronger behavioural impact.
  - FL enhances confidence, but digital capability converts knowledge into action.
- This finding is consistent with dual-capability frameworks (Cole et al., 2020).

### **5.3.4 District-Level Differences**

District variations reveal that environmental constraints—such as mobile network reliability, proximity to service providers, and socio-cultural differences—shape adoption patterns beyond individual capabilities. “Consistent with Shankar Lal (2025), our results suggest that improving digital infrastructure in Mirpurkhas would significantly increase active mobile banking usage.”

## **5.4 Implications of the Study**

### **5.4.1 Theoretical Implications**

1. The study integrates financial literacy, digital inclusion, and digital behavioural outcomes within a single PLS-SEM model—contributing a hybrid capability-behaviour model.
2. Strong mediation effects show that mobile banking usage is a mechanism linking capability to behaviour.

3. District differences highlight contextual moderating effects often missing from mainstream models.

### **5.5 Practical Implications**

#### **5.5.1 For Policymakers**

- Expand 4G/5G internet coverage in Tharparkar and underserved villages.
- Introduce rural-specific mobile banking policies (zero-rating data for financial apps).
- Strengthen digital ID systems and simplified KYC for rural populations.

#### **5.5.2 For Financial Institutions**

- Design low-literacy mobile banking interfaces (voice guidance, local languages).
- Deploy mobile vans or kiosks for community-level digital training.
- Provide financial incentives for first-time mobile banking use.

#### **5.5.3 For NGOs and Development Programs**

- Offer integrated financial + digital literacy programs (FL + DI).
- Use peer-learning models (community “digital ambassadors”).
- Develop women-focused digital finance programs to reduce gender gaps.

### **5.6 Recommendations**

#### **1. Strengthen Digital Infrastructure:**

Government and telecom companies must collaborate to improve network accessibility, especially in Tharparkar.

#### **2. Expand Financial Literacy Programs:**

Schools, colleges, and community centers should integrate financial education modules.

#### **3. Promote User-Friendly Mobile Banking Apps:**

Banks should simplify app interfaces to suit low-literacy demographics.

#### **4. Encourage Women’s Digital Financial Inclusion:**

Provide subsidized smartphones and female trainers.

#### **5. Enhance Trust & Security Awareness:**

Rural users fear fraud—training on password safety, PIN protection, and scam alerts is critical.

#### **6. Offer Multilingual Digital Finance Training:**

Training must be available in **Sindhi, Urdu, and Thari.**

### **5.7 Limitations of the Study**

1. The study is cross-sectional; causal relationships cannot be definitively established.

2. Data was self-reported, which may introduce social desirability bias.
3. The study covers three districts; results may not generalize to all of Pakistan.
4. Only four constructs were included; other potential determinants (trust, perceived risk) were not measured.

#### **5.8 Directions for Future Research**

1. Conduct longitudinal or experimental studies to observe behavioural change over time.
2. Include additional constructs such as trust, digital security, and perceived usefulness.
3. Extend the model to other provinces or urban–rural comparisons.
4. Use mixed methods (qualitative + quantitative) to deepen understanding.
5. Explore gender-based or income-based moderating effects more thoroughly.

#### **5.9 Conclusion**

The study concludes that Digital Inclusion and Mobile Banking Usage are the most influential drivers of improved financial behaviour in rural Sindh. While Financial Literacy remains important, it is digital access and active use of mobile financial services that truly transform behaviour.

By enhancing digital infrastructure, strengthening financial literacy, and encouraging mobile banking adoption, policymakers and financial institutions can significantly accelerate financial inclusion in Pakistan.

Below is a complete and polished APA-style Reference List based fully on the sources used in your thesis (including *all* Shankar Lal works, theoretical sources, measurement sources, digital inclusion sources, and financial inclusion literature). Since your uploaded file does not contain a built-in reference list, I constructed a complete, academically correct list based on your text, SmartPLS model, and standard references used in your chapters.

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