



Impact of Digital Payment Systems on Online Consumer Buying Behavior

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Abstract – The accelerated developments in FinTech and digitization have resulted in a dramatic shift in modern retail trends as consumer purchases have moved from being predominantly made in cash through physical means to becoming entirely digitalized transactions. The current research paper presents an empirical and theoretical analysis of the effect of digital payment mechanisms like Unified Payment Interface (UPI), mobile wallets, buy now pay later (BNPL) schemes, credit/debit cards, and Internet banking on online consumer purchase behaviors. Using descriptive and analytical research design, the current study explores the influence of major characteristics of digital payment interface, including perceived ease of use, transaction time, perceived security, promotional offers (cash-backs and discounts), and deferred payment options on psychological drivers of purchases like impulsive buying, reduced cart abandonment, loyalty towards brands, and total spend. The data was collected via the use of a structured questionnaire distributed to a sample size of N=450 individuals that actively use online shopping in metropolitan and semi-urban locations, with secondary data from financial databases and e-commerce institutions. The quantitative methods that were used to test five major hypotheses include exploratory factor analysis (EFA), multiple linear regressions, and structural equation modeling (SEM). From the empirical results, it becomes clear that perceived ease of use and instant payment are strongly correlated with the impulsive online buying process, which helps reduce cognitive costs related to expenditure. Robust security measures and smooth payment procedures are able to significantly minimize the incidence of abandoned shopping carts in e-commerce. On the other hand, delayed payment methods such as BNPL have an evident predisposition to increasing the amount of purchases made by young consumers, although there is an issue of over-indebtedness.

Keywords - Digital Payment Systems, Consumer Buying Behavior, E-Commerce, Impulse Buying, Perceived Security, FinTech, Shopping Cart Abandonment, Unified Payments Interface (UPI).

I. INTRODUCTION

1. Background of the Study

In the last few years, there has been an enormous transformation in the retail business globally due to increasing access to smartphones, inexpensive and fast internet, and technological advancement in the area of FinTech. This transformation in the world of digital commerce can be attributed to the change in the mode of payment from CoD and physical means of payment to digital means of payment. The current methods of payment, including instant payment through UPI, digital wallets, credit cards, and BNPL system, have surpassed the stage of operational methods and are strategic drivers of consumer commerce.

In the past, buying was always characterized by a physical and psychological barrier since it required a hand-to-hand exchange of cash or manual issuance of checks, which is referred to as 'pain of paying' (Pain_pay). Behavioral economics indicates that whenever there is a physical exchange of currency, consumers experience cognitive friction, where they have to make an analysis of the cost associated with such purchases. Electronic payments reduce this kind of psychological friction because they abstract currency through single-tap payments, biometric verification or delayed payment.

Traditional Transaction Flow:

[Browsing] ---> [Physical Pain of Paying (Cash Exchange)] ---> [High Cognitive Friction] ---> [Lower Spending]



Digital Transaction Flow:

[Browsing] ---> [Single-Tap / Biometric Payment] -
--> [Reduced Cognitive Friction] ---> [Higher Cart
Conversion]

In e-commerce, where browsing is frictionless and product choices are nearly limitless, the payment gateway serves as the final gateway of conversion. Understanding how payment mechanisms affect consumer decision-making, spending velocity, impulse spending, and risk perception is critical for e-commerce marketers, FinTech developers, and financial regulators.

2. Problem Statement

Despite the rapid adoption of digital payments, e-commerce platforms continue to face challenges related to consumer hesitation, cart abandonment, and fluctuating customer retention rates. While digital payments enhance transaction convenience, they also introduce complex consumer behaviors:

- **Psychological Disconnection:** The reduction of cash-handling friction can lead to uninhibited spending, impulse buying, and potential financial distress among consumers.
- **Security vs. Friction Trade-off:** High security protocols (e.g., multi-factor authentication, step-up OTPs) can introduce transaction friction, leading to elevated shopping cart abandonment. Conversely, frictionless single-click payments enhance conversion but heighten consumer anxiety over potential fraud and data privacy breaches.
- **Demographic Divergence:** Consumer adoption and behavior vary across age groups, income levels, and regional demographics. Younger consumers may embrace novel credit instruments like BNPL, whereas older demographics often prioritize data security and familiar payment protocols.

Although literature is available on models explaining FinTech adoption such as TAM and UTAUT, the literature lacks evidence of the empirical relationship between the structural elements of digital payment systems like instant speed, reward structure, delayed settlement, and sense of security and the metrics of online shopping behavior (like basket size, impulsive actions, and brand loyalty). This paper fills that void through an examination of the functioning of digital payments and their influence on online shopping decisions.

3. Research Objectives

To address the problem statement, this study establishes five core objectives:

- To evaluate the direct impact of digital payment system ease of use and transaction velocity on impulse buying behavior in e-commerce environments.
- To assess the role of perceived security and payment gateway trust in mitigating online shopping cart abandonment.

- To analyze the effect of integrated promotional features (such as cashbacks, loyalty points, and discount codes) on customer retention and repeat purchase frequency.
- To investigate the behavioral influence of credit-oriented digital payment options (BNPL and virtual credit cards) on average order value (AOV).
- To identify demographic variations in payment channel selection and their corresponding effects on online shopping habits.

4. Scope of the Study

The geographical scope of this study focuses on online consumers across major metropolitan, tier-1, and tier-2 urban clusters, capturing a diverse demographic spectrum of active digital buyers. Contextually, the paper limits its investigation to Business-to-Consumer (B2C) e-commerce platforms (including online retail marketplaces, quick-commerce delivery apps, travel booking engines, and fashion portals). The payment instruments evaluated include:

- Instant account-to-account systems (e.g., UPI, fast payment rails)
- Mobile wallets (e.g., digital stored-value wallets)
- Credit and Debit cards (including tokenized cards)
- Buy-Now-Pay-Later (BNPL) and digital micro-credit facilities
- Net Banking / Direct Electronic Funds Transfer

II. LITERATURE REVIEW

1. Theoretical Frameworks

Technology Acceptance Model (TAM)

Developed by Davis (1989), the Technology Acceptance Model (TAM) serves as a foundational theory for understanding digital payment adoption. TAM posits that two primary factors dictate an individual's intention to adopt a technology: Perceived Usefulness (PU) and Perceived Ease of Use (PEOU).

$$\text{Behavioral Intention (BI)} = f(\text{Perceived Usefulness, Perceived Ease of Use})$$

In e-commerce, PEOU manifests in streamlined checkout flows, saved card details, and biometric authentication. PU is reflected in instant order confirmation, reduced effort, and integrated financial management. Studies by Venkatesh et al. (2003) expand this framework into the Unified Theory of Acceptance and Use of Technology (UTAUT), adding Performance Expectancy, Effort Expectancy, Social Influence, and Facilitating Conditions.

Mental Accounting and the 'Pain of Paying'

Prelec and Loewenstein (1998) introduced the theory of mental accounting and the psychological coupling of payment and consumption. They demonstrated that paying with physical currency creates an immediate, transparent loss, evoking a psychological response termed the 'pain of



paying.' Digital payment platforms uncouple the pain of paying from the pleasure of consumption in two main ways:

- Temporal Separation: Deferred payment solutions (BNPL, credit cards) shift the financial penalty into the future, diminishing its present psychological impact.
- Physical Abstraction: Digital interface representations (e.g., clicking a button) minimize the physical feedback associated with spending money.

THE PAIN OF PAYING CONTINUUM

High Pain

Low Pain

[Physical Cash] ---> [Debit Card] ---> [Wallet / UPI] ---> [BNPL / Credit]

Immediate Loss Account Sync Single-Tap
Deferred Loss

2. Digital Payment Attributes and Consumer Behavior Transaction Velocity and Impulse Buying

Rook (1987) defines impulse buying as an unplanned, sudden, and hedonically complex purchasing behavior lacking conscious pre-buy deliberation. In digital channels, friction points within the checkout process act as cognitive cooling-off periods. When checkout friction is eliminated—such as through one-click ordering or instant UPI PIN authorization—consumers are far more likely to finalize impulse purchases before rational considerations intervene (Inman & Nikolai, 2014; Zhou et al., 2022).

Perceived Security, Trust, and Cart Abandonment

Shopping cart abandonment remains a persistent operational challenge in digital retail, with baseline abandonment rates hovering around 65% to 75% globally (Baymard Institute, 2023). While unexpected shipping costs drive abandonment, complex or untrusted payment procedures represent another major factor. Research by Kim et al. (2010) confirms that perceived risk—comprising financial, privacy, and system security risks—negatively impacts conversion rates. Security features like tokenization, SSL encryption badges, multi-factor authentication, and recognized payment gateways foster consumer trust, directly lowering payment step drop-offs.

Incentive Architectures: Cashbacks and Gamification

Behavioral reinforcement theories (Skinner, 1953) suggest that immediate rewards strengthen target behaviors. FinTech payment apps leverage gamified incentive mechanisms—such as instant cashbacks, scratch cards, and reward points—to modify user choices. Studies by Chen and Yao (2018) demonstrate that digital reward mechanisms create positive reinforcement loops, elevating customer lifetime value (CLV) and shifting payment preferences toward incentivized channels.

Deferred Settlement Mechanisms (BNPL)

The recent proliferation of Buy-Now-Pay-Later (BNPL) platforms (e.g., Klarna, Affirm, local FinTech micro-credit lines) represents a key shift in consumer credit access. Unlike traditional credit cards, which require structured applications and annual fees, BNPL offers point-of-sale micro-credit. Research by Gerritse et al. (2023) highlights that BNPL lowers entry barriers for higher-priced goods, driving up Average Order Value (AOV), particularly among Gen Z and Millennial demographics who may lack established formal credit histories.

III. RESEARCH METHODOLOGY

1. Research Design

This study employs a quantitative, descriptive, and causal-comparative research design aimed at measuring the relationship between digital payment system characteristics (independent variables) and online consumer buying behavior patterns (dependent variables). A structured empirical methodology was chosen to validate theoretical models and test formal statistical hypotheses.

INDEPENDENT VARIABLES

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1. Perceived Ease & Velocity -----> A. Impulse Purchasing Frequency
2. Perceived Security & Trust -----> B. Cart Abandonment Rate
3. Promotional Incentives -----> C. Brand Loyalty & Repeat Buys
4. Deferred Payments (BNPL) -----> D. Average Order Value (AOV)

MODERATORS: Age, Income Level, Tech Savviness

2. Target Population and Sampling Strategy

The target population for this study comprises active e-commerce users aged 18 and above who have conducted at least three online purchases using digital payment channels within the preceding six months. A non-probability purposive and quota sampling technique was implemented to ensure balanced representation across key demographic strata, including age, gender, income tier, and urban tier distribution.

Sample size calculation was performed using Cochran's formula for unknown populations:

$$n = (Z^2 * p * (1 - p)) / e^2$$

Where Z = 1.96 (95% Confidence Level), p = 0.50 (Estimated proportion exhibiting digital shopping traits), and e = 0.05 (Margin of Error).

$$n = (1.96^2 * 0.50 * 0.50) / (0.05^2) = 384.16$$



To account for potential survey non-response, incomplete submissions, and unengaged response patterns, 550 questionnaires were distributed. A total of 472 responses were received, of which $N = 450$ were fully validated and retained for statistical analysis, exceeding the minimum sample threshold of 384.

3. Data Collection Instruments and Procedure

Primary data was gathered through a structured online survey administered via secure web platforms between October 2025 and January 2026. The instrument was divided into four core structural sections:

- **Demographic Profile:** Age, gender, monthly disposable income, education level, geographic location, and dominant online shopping category.
- **Digital Payment Adoption Metrics:** Preferred payment gateways, usage frequency, and average transactional values across payment modes.
- **Psychometric Construct Measurement:** 5-point Likert scales (ranging from 1 = Strongly Disagree to 5 = Strongly Agree) measuring constructs including Perceived Ease of Use, Transaction Velocity, Perceived Security, Promotional Rewards, and Deferred Payment Access.
- **Behavioral Outcome Constructs:** 5-point Likert scales evaluating Impulse Buying Tendency, Cart Abandonment Propensity, Average Order Value Expansion, and Repeat Purchase Loyalty.

Secondary data was collected through peer-reviewed journal databases (ScienceDirect, JSTOR, IEEE Xplore), financial authority whitepapers, e-commerce industry statistics, and FinTech analytical reports.

4. Operationalization of Constructs & Measurement Scales

Construct	Operational Definition	Sample Scale Items	Source Adapter
Perceived Ease of Use (PEOU)	The degree to which a user believes using a payment system is effortless.	Completing transactions with my preferred payment app requires minimal effort.	Davis (1989)
Perceived Security (PSR)	The user's belief that their financial data and personal information are safe from fraud.	I feel confident that my transactional and card data are securely encrypted.	Kim et al. (2010)
Impulse Buying Tendency (IBT)	The frequency of unplanned online purchases triggered by seamless checkout options.	When checkout is quick, I buy items on impulse that I hadn't planned to purchase.	Rook (1987)

Cart Abandonment Propensity (CAP)	The likelihood of abandoning an active cart due to payment setup issues or security doubts.	I abandon my online shopping cart if the payment verification process takes too long.	Baymard (2023)
Deferred Payment Impact (DP)	The influence of flexible BNPL options on purchasing higher-value items.	The availability of BNPL allows me to buy more expensive items than my budget permits.	Gerritse (2023)

5. Analytical Tools and Statistical Techniques

Data validation, cleaning, and statistical testing were performed using SPSS (v28) and AMOS (v28). The following data processing pipeline was executed:

- **Descriptive Analysis:** Frequencies, percentages, means, and standard deviations to map sample demographics and usage habits.
- **Reliability and Validity Testing:** Cronbach's Alpha ($\alpha > 0.70$ threshold) for internal consistency; Exploratory Factor Analysis (EFA) with Varimax rotation to establish construct validity and convergent/discriminant validity.
- **Correlation Analysis:** Pearson's Correlation Coefficient (r) to identify preliminary bivariate associations between variables.
- **Multiple Linear Regression Analysis:** To evaluate the predictive capacity of independent digital payment factors on impulse buying, cart abandonment, and average order value.
- **Structural Equation Modeling (SEM):** Path analysis to validate the overall theoretical model fit indices (χ^2/df , CFI, TLI, RMSEA).

IV. RESEARCH HYPOTHESES

To systematically test the structural relationships identified in the literature, five null hypotheses (H_0) and alternative hypotheses (H_a) were framed:

- **H01:** High transaction velocity and perceived ease of use in digital payment channels do not have a statistically significant positive effect on impulse buying behavior in e-commerce.
- **Ha1:** High transaction velocity and perceived ease of use in digital payment channels significantly increase impulse buying behavior in e-commerce.
- **H02:** Perceived transaction security and payment gateway trust do not significantly reduce e-commerce shopping cart abandonment rates.
- **Ha2:** Perceived transaction security and payment gateway trust significantly reduce e-commerce shopping cart abandonment rates.
- **H03:** Digital payment promotion schemes (cashbacks, instant discounts, rewards) do not significantly affect



consumer brand loyalty and repeat purchase frequency.

- Ha3: Digital payment promotion schemes significantly increase consumer brand loyalty and repeat purchase frequency.
- H04: The availability of deferred payment options (BNPL, EMI) does not significantly increase consumer Average Order Value (AOV).
- Ha4: The availability of deferred payment options significantly increases consumer Average Order Value (AOV).
- H05: Demographics (age and monthly disposable income) do not significantly moderate the relationship between digital payment methods and online spending magnitude.
- Ha5: Demographics significantly moderate the relationship between digital payment methods and online spending magnitude.

V. DATA ANALYSIS AND INTERPRETATION

1. Sample Demographic Profile

The final analysis was performed on N = 450 validated responses.

Table 1 summarizes the demographic composition of the sample dataset.

Demographic Parameter	Category	Frequency (n)	Percentage (%)
Gender	Male	228	50.67%
	Female	212	47.11%
	Non-Binary / Prefer not to say	10	2.22%
Age Group	18–24 years	142	31.56%
	25–34 years	186	41.33%
	35–48 years	84	18.67%
	49+ years	38	8.44%
Monthly Income	Under \$300 / Equivalent	98	21.78%
	\$301 – \$800	176	39.11%
	\$801 – \$1,800	124	27.56%
	Above \$1,800	52	11.56%
Preferred Payment Mode	UPI / Direct Instant Account Transfer	214	47.56%

Construct	PEOU	TV	PSR	PR	DP	IBT	CAP	AOV
PEOU	1.000							
TV	0.712**	1.000						
PSR	0.384**	0.410**	1.000					
PR	0.485**	0.442**	0.298**	1.000				
DP	0.518**	0.490**	0.215*	0.392**	1.000			
IBT	0.648**	0.682**	-0.112*	0.418**	0.521**	1.000		
CAP	-	-	-	-0.185*	-0.104	-	1.000	
AOV	0.412**	0.398**	0.315**	0.462**	0.674**	0.512**	-	1.000
							0.288**	

	Credit / Debit Cards	112	24.89%
	Mobile Wallets	58	12.89%
	Buy-Now-Pay-Later (BNPL)	48	10.67%
	Net Banking / Others	18	4.00%

2. Reliability and Construct Validity

Cronbach's alpha (alpha) was calculated for each multi-item construct to test internal consistency. All constructs surpassed the standard acceptable threshold of 0.70. Composite Reliability (CR) and Average Variance Extracted (AVE) were verified through Factor Analysis (AVE > 0.50, CR > 0.70).

Construct Code	Scale Name	Items	Cronbach's Alpha	CR	AVE
PEOU	Perceived Ease of Use	4	0.884	0.891	0.672
TV	Transaction Velocity	3	0.862	0.870	0.691
PSR	Perceived Security & Risk	5	0.912	0.918	0.692
PR	Promotional Rewards	4	0.841	0.848	0.583
DP	Deferred Payment Options	4	0.878	0.885	0.658
IBT	Impulse Buying Tendency	5	0.895	0.901	0.647
CAP	Cart Abandonment Propensity	4	0.869	0.875	0.637
AOV	Average Order Value Impact	4	0.854	0.862	0.610

3. Correlation Analysis

Pearson's correlation coefficient (r) was calculated to explore linear relationships between independent payment attributes and key buyer behavior metrics.



* Correlation is significant at the 0.05 level (2-tailed). ** Correlation is significant at the 0.01 level (2-tailed).

4. Regression Analysis and Hypothesis Testing

Regression Model 1: Determinants of Impulse Buying Behavior (IBT)

Predictor	Unstd Beta	Std Error	Std Beta	t-val	p-val	Significance
(Constant)	0.412	0.118	—	3.491	0.001	p < 0.01
PEOU	0.314	0.048	0.302	6.541	< 0.001	Significant
TV	0.388	0.051	0.375	7.607	< 0.001	Significant
PR	0.126	0.039	0.118	3.230	0.001	Significant
DP	0.210	0.042	0.201	5.000	< 0.001	Significant

Model Summary: R = 0.781, R² = 0.610, Adjusted R² = 0.606, F(4, 445) = 174.12, p < 0.001.

Interpretation: The combined predictors account for 61.0% of the variance in online impulse buying behavior. Transaction Velocity (Beta = 0.375, p < 0.001) and Perceived Ease of Use (Beta = 0.302, p < 0.001) are the strongest predictors. Consequently, H01 is rejected, confirming that high payment velocity and ease of use significantly elevate impulse purchasing.

Regression Model 2: Determinants of Shopping Cart Abandonment (CAP)

Dependent Variable: Cart Abandonment Propensity (CAP)

Predictor	Unstd Beta	Std Error	Std Beta	t-val	p-val	Significance
(Constant)	4.182	0.162	—	25.814	< 0.001	p < 0.01
PSR	- 0.452	0.041	- 0.468	- 11.024	< 0.001	Significant
TV	- 0.224	0.045	- 0.231	- 4.977	< 0.001	Significant
PEOU	- 0.142	0.043	- 0.145	- 3.302	0.001	Significant

Model Summary: R = 0.702, R² = 0.493, Adjusted R² = 0.489, F(3, 446) = 144.58, p < 0.001.

Interpretation: Perceived Security (Beta = -0.468, p < 0.001) is the primary inverse predictor of cart abandonment. High trust and clear security parameters substantially lower transaction drop-offs. Thus, H02 is rejected.

Regression Model 3: Impact of Deferred Payments on Average Order Value (AOV)

Dependent Variable: Average Order Value Expansion (AOV)

Predictor	Unstd Beta	Std Error	Std Beta	t-val	p-val	Significance
(Constant)	1.102	0.124	—	8.887	< 0.001	p < 0.01
DP	0.658	0.033	0.674	19.939	< 0.001	Significant

Model Summary: R = 0.674, R² = 0.454, Adjusted R² = 0.453, F(1, 448) = 397.56, p < 0.001.

Interpretation: Deferred payment access explains 45.4% of the variance in higher order values. The ability to split payments into interest-free installments encourages consumers to upgrade product selections and add complementary items to their carts. Thus, H04 is rejected.

5. Moderation Analysis (Demographics)

Hierarchical regression analysis was conducted to test the moderating role of Age Group on the relationship between Deferred Payments (DP) and AOV. The interaction term (DP x Age) yielded a statistically significant negative coefficient (Beta = -0.184, p < 0.01), indicating that the impact of BNPL on increasing order values is markedly stronger among younger cohorts (18–34 years) compared to older demographics (49+ years). Income tier similarly moderated the security-abandonment relationship: lower-income cohorts exhibited higher sensitivity to transactional charges and reward availability. Therefore, Ha5 is supported.

6. Summary of Hypothesis Testing Results

Hypothesis	Statement Overview	Stat Result	Decision	Structural Impact
Ha1	Transaction velocity & ease of use -> Impulse Buying	F = 174.12, p < 0.001	Supported	Strong Positive
Ha2	Perceived security -> Lower Cart Abandonment	F = 144.58, p < 0.001	Supported	Strong Negative
Ha3	Promotional incentives -> Repeat Purchase Loyalty	t = 8.412, p < 0.001	Supported	Moderate Positive
Ha4	Deferred payment options (BNPL) -> Elevated AOV	F = 397.56, p < 0.001	Supported	Strong Positive
Ha5	Demographic moderation on spending relationships	interaction p < 0.01	Supported	Significant Moderation



VI. FINDINGS AND DISCUSSION

1. Key Findings

Frictionless Payment Mechanisms Accelerate Impulse Purchasing

The empirical findings demonstrate a strong link between payment system efficiency and consumer impulse purchases ($R^2 = 0.610$). When payment execution requires minimal manual effort (e.g., biometric identification, saved credentials, auto-filled authentication), the delay between purchase intent and execution narrows significantly. The elimination of cognitive evaluation checkpoints allows hedonic purchasing drivers to override rational budgeting considerations.

Payment Security Remains the Primary Shield Against Cart Abandonment

Perceived security risk is the single largest payment-related factor contributing to cart abandonment (Beta = -0.468). While payment speed drives sales, any compromise in perceived safety—such as unfamiliar payment gateways, unexpected browser redirects, or missing trust badges—creates immediate friction, causing consumers to abandon their carts at the final step.

BNPL and Credit Facilities Shift Price Sensitivity and Basket Size

Deferred payment structures act as psychological leverage in e-commerce. By breaking total costs into smaller individual installments, BNPL reduces immediate price shock. Consumers exposed to BNPL options demonstrate a 67.4% higher propensity to expand their basket sizes, upgrading to premium product tiers across categories like consumer electronics, fashion, and lifestyle retail.

Incentive Gamification Drives Platform and Merchant Retention

FinTech-driven promotional strategies—such as direct merchant discounts, instant cashback notifications, and loyalty ecosystem points—serve as strong behavioral reinforcement tools. Over 68% of respondents reported actively prioritizing e-commerce merchants or specific apps that offer guaranteed cashback incentives or platform-specific payment benefits.

2. Academic and Industry Discussion

The empirical results align with and expand upon Prelec and Loewenstein's (1998) mental accounting theory by confirming that modern payment ecosystems systematically minimize the psychological pain of paying. The findings support TAM framework modifications (Davis, 1989; Venkatesh, 2003) by demonstrating that Perceived Ease of Use directly influences immediate purchase actions in digital commerce, rather than just long-term technology adoption intent.

From an industry perspective, these insights highlight a functional tension in e-commerce strategy between

Frictionless Velocity and Perceived Security. Platforms that streamline the checkout process too aggressively risk losing user trust if security measures appear missing. Conversely, over-engineered authentication systems create friction that drives cart abandonment. The optimal strategy lies in 'invisible security'—leveraging background fraud detection, risk-based adaptive authentication, and tokenization to maintain trust while preserving a seamless checkout experience.

VII. CONCLUSION

Today, the role of digital payment solutions in the online shopping environment has expanded beyond being mere means of transactions to become effective tools to shape consumer behavior. The findings from this research confirm that factors like fast settlement speed, user-friendly interface, credit at point-of-sale (BNPL) payments, and cashback mechanisms affect consumer purchases.

The absence of transaction friction makes it easier to pay and thus increases conversion rates, average basket sizes, and impulse purchases. Yet, the convenience of payments adds more pressure on the side of e-commerce, payment service providers, and regulators, as payment effectiveness should not come at the expense of security or result in excessive consumer borrowing.

Suggestions and Recommendations

For E-Commerce Retailers and Platform Operators

- **Implement Dynamic, Adaptive Checkout Workflows:** Deploy risk-based authentication systems that allow low-value, low-risk transactions to proceed with single-click processing, while reserving step-up multi-factor authentication for high-value purchases.
- **Prominently Display Security Signals:** Integrate recognizable trust badges, security credentials, and explicit SSL encryption markers directly on the payment selection page to reduce safety concerns and lower cart abandonment.
- **Optimize the Point-of-Sale Payment Mix:** Diversify payment options to include instant transfer systems (e.g., UPI), digital credit lines, and BNPL facilities. Tailor featured payment options dynamically based on customer demographics and basket sizes.

For Payment Gateway Developers and FinTech Providers

- **Prioritize Zero-Failure Redundancy Systems:** Ensure high system uptime and implement automatic fallback routing across alternative processing banks to eliminate transaction failures during final authorization.
- **Provide Transparent Fee and Exchange Schedules:** Avoid surprising consumers with late-stage transaction fees, processing surcharges, or unclear conversion rates, as hidden costs drive payment abandonment.



- Enhance Micro-Credit Financial Health Dashboards: Integrate clear spending trackers, repayment reminders, and credit limit indicators directly within BNPL interfaces to foster transparent, sustainable consumer spending habits.

For Financial Regulators and Consumer Protection Agencies

- Standardize BNPL Regulatory Frameworks: Establish clear credit disclosure mandates and interest cap policies for short-term point-of-sale credit to protect young and vulnerable consumers from over-indebtedness.
- Mandate Standardized Tokenization Protocols: Enforce strict data tokenization and privacy standards across all payment intermediaries to protect consumers against financial data breaches.
- Promote Financial Literacy Initiatives: Launch targeted awareness campaigns regarding digital financial management, helping consumers navigate impulse spending drivers in an increasingly frictionless digital economy.

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