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3 **Consumer Satisfaction and Preference toward Blinkit versus Traditional Grocery Stores: An Empirical**
4 **Study in Bhopal City.**
5

6 **Abstract:**

7 The explosive rise of quick commerce platforms like Blinkit has had a profound effect on how we purchase
8 groceries in cities across emerging markets (Chakraborty et al., 2022). This research investigates how consumers
9 have adopted q-commerce, their levels of satisfaction with this platform, and its competition with traditional Kirana
10 stores in Bhopal (Kaur et al., 2021). A total of 202 urban residents participated in this study's cross-sectional,
11 quantitative approach to collect primary data. When the dataset was analyzed again, it had an overall high
12 satisfaction index regarding the q-commerce platform (94.85%). Using a multiple linear regression model to
13 examine predictors of consumer satisfaction, product availability ($\beta=0.875$, $p<0.001$) and convenience ($\beta=0.315$,
14 $p<0.001$) were found to be the strongest predictors of overall consumer satisfaction with q-commerce; pricing was
15 found to be negatively correlated with overall consumer satisfaction ($\beta=-0.200$, $p<0.001$). Additionally, 51% of
16 respondents reported decreased visits to Kirana stores. These results indicate that traditional grocery retailers will
17 need to develop localized digital delivery options to overcome the structural disadvantages created by q-commerce
18 platforms (Al Amin et al., 2021).
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20 **Key words:-**

21 Quick Commerce; Consumer Satisfaction; Retail Transformation; Multiple Regression; Kirana Stores; Blinkit
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26 **Introduction**

27 Understanding the Indian retail ecosystem landscape since 2020 had never been so complicated; however, together
28 with the rapid digitization of consumer goods and the rise of q-commerce (Ray et al., 2019) since 2020, this journey
29 continued. The most significant development has been the formation of mobile delivery services that guarantee
30 delivery of orders within a few minutes (Zhao and Bacao, 2020). For example, Blinkit, by utilizing well-built digital
31 interfaces and technologically advanced last-mile delivery options, has risen to become one of the major players in
32 the market. In contrast to the traditional, independent Kirana stores that have been the backbone of Indian
33 neighborhoods for many years (Grewal et al., 2020), the small independent retailers that established themselves in
34 their communities through their close relationship with their customers, their physical location, their relationships
35 with customers and their use of informal credit networks for supplying goods to customers have always provided
36 resistance to market forces. Today's rapid rise of q-commerce will disrupt this. Moreover, extrapolating consumer
37 buying behavior from mega cities into Tier-2 cities such as Bhopal does not yield conclusive evidence, thus raising
38 the fundamental research question posed in this study: what are the true determinants of customer satisfaction, and
39 how does the structural change created by q-commerce affect customer dependency on Kirana stores? (Dash et al.,
2021).

40 **Literature Review**

41 Transition from traditional retailing to quick commerce: Retailers switched from having set shipping dates for e-
42 commerce to now offering same-day delivery (quick commerce). The predominant theme of today's discussions is
43 hyper-local delivery networks, which disrupt the supply chain and create new challenges for traditional retail
44 channels. In particular, quick delivery offers heightened expectations from consumers (Eger et al.; 2021).

45 Technology acceptance & factors influencing satisfaction: This research paper uses the Technology Acceptance
46 Model (TAM) as its theoretical framework. The two constructs of PU (Perceived Usefulness) & PEOU (Perceived
47 Ease of Use) respective measure speed, availability, application user interface/user experience (UI/UX)
48 (Bauerová&Klepek, 2018; Troise et al., 2021). Moreover, based on recent academic contributions, research shows
49 that Gen Z & Millennials prioritize instant gratification over strict price-driven value based on their digital-native
50 backgrounds (Kaur et al.; 2021).

51 Resilience & vulnerability of institutions: While digital channels continue to grow, retailers using a conventional
52 retailing model continue to hold significant market share based on the relational capital of existing customers.

53 Traditional retailers are experiencing structural technology gaps & outdated last mile delivery as their biggest risks
54 to continued success. Researchers have recommended that independent grocers establish hybrid digital business
55 models in order to survive (Wang & Somogyi, 2018).

56 The current body of research mainly investigates the use of rapid delivery services in urban areas (Tier 1), while
57 little research has been conducted on how these disruptive technologies emerge in rapidly growing Tier 2 cities.
58 Additionally, it is well known that conventional brick-and-mortar retailers are becoming more susceptible to rapid
59 delivery services, but comparatively few studies have quantitatively assessed what service levels are sufficient for
60 consumers to switch their loyalty from local Kirana retailers to q-commerce companies. Thus, this research will
61 attempt to fill that void by examining the relationship between digital hyper-convenience and traditional relational
62 capital in relation to Bhopal using the contextual sociocultural conditions of the city.

63 **Research Objectives**

64 The primary purpose of this empirical study is to evaluate the shifting paradigm of grocery retailing in Bhopal City
65 by analyzing consumer adoption of quick commerce platforms compared to traditional physical stores. Specifically,
66 the study aims:

- 67 1). To measure the overall level of consumer satisfaction with quick commerce delivery platforms among digitally
68 active urban residents.
- 69 2). To evaluate the impact of key service dimensions—specifically product availability, convenience, delivery speed,
70 and pricing—on overall consumer satisfaction.
- 71 3). To determine whether demographic variables, such as age cohort and occupational status, create significant
72 variance in the adoption and usage frequency of q-commerce services.
- 73 4). To assess the systemic impact of quick commerce adoption on consumer visitation frequency and purchasing
74 behavior at traditional neighborhood Kirana stores.

76 **Research Hypotheses**

77 Drawing upon the theoretical framework of technology acceptance and the identified service dimensions, the
78 following hypotheses were formulated for empirical testing to understand consumer behavior toward quick
79 commerce platforms:

- 80 H1: Delivery speed has a significant positive influence on overall consumer satisfaction with quick commerce
81 platforms.
- 82 H2: Platform convenience has a significant positive influence on overall consumer satisfaction with quick
83 commerce platforms.
- 84 H3: Product availability has a significant positive influence on overall consumer satisfaction with quick
85 commerce platforms.
- 86 H4: Pricing has a significant negative influence on overall consumer satisfaction with quick commerce
87 platforms.
- 88 H5: There is a significant variance in the adoption of, and overall satisfaction with, quick commerce platforms
89 across different age cohorts.
- 90 H6: Occupational status significantly dictates the usage frequency of quick commerce services.

92 **Research Methodology**

93 **Research Design and Sampling Procedure:** The study adopted a descriptive and cross-sectional approach
94 (quantitative research). The target population of this research was digitally active individuals residing in urban areas
95 of Bhopal. Convenience sampling was used to obtain a sample of N=202 that met the required criteria for the study
96 as specified by Dash et al (2021).

97 **Data Collection Tool:** Data were collected by means of a structured self-administered questionnaire containing
98 demographic questions, behavioural measures of service usage, and rating of service quality on five dimensions:
99 speed; convenience; availability; price; and overall satisfaction each measured on a 5-point Likert scale (1 = strongly
100 disagree; 5 = strongly agree) (Tandon et al., 2018).

101 Data Analysis Procedures: The data were analysed using Python statistical software (Pandas, SciPy, Statsmodels);
 102 descriptive statistics provided summary statistics (aspects of the distribution of each variable) and assessed the
 103 distributional characteristics of the variables. To assess the reliability of the scales (serving as outcomes of interest)
 104 in the study, Cronbach's alpha, composite reliability (CR), and average variance extracted (AVE) were calculated.
 105 The inferential analyses included Pearson correlations; analyses of variance (ANOVA); independent samples t-test;
 106 and ordinary least squares (OLS) multiple regression. Multicollinearity among the independent variables was
 107 assessed using the variance inflation factor (VIF) (Hair et al., 2019).

108 As part of our Preliminary Data Analysis, initial screening was done on all raw data. The purpose of this screening is
 109 to identify and remediate any missing or largely unengaged responses to ensure that the integrity of the structure of
 110 the final dataset is maintained. Additionally, a pilot study was conducted using a small number (< 10) from target
 111 population to determine whether the instrument used had face validity, was clear, and demonstrated a logical flow
 112 before the full implementation. Statistical hypothesis testing was conducted at a 95% level of confidence ($p < 0.05$).
 113 Ethical guidelines have been maintained throughout this investigation (e.g., all individuals provided informed
 114 consent) and anonymizing participant personally identifiable information helps reduce any social desirability bias
 115 that may exist and increases the likelihood of participants providing objective responses.

116 Results

117 Demographic Characteristics

118 The sample consisted largely of young, digitally active consumers: 63.37% were aged 18–25, and 74.26% identified
 119 as students. The majority of respondents (83.17 per cent) had awareness of q-commerce platforms, with ‘fast
 120 delivery’ (36.14 per cent) being the key adoption driver – (Ray et al., 2019).

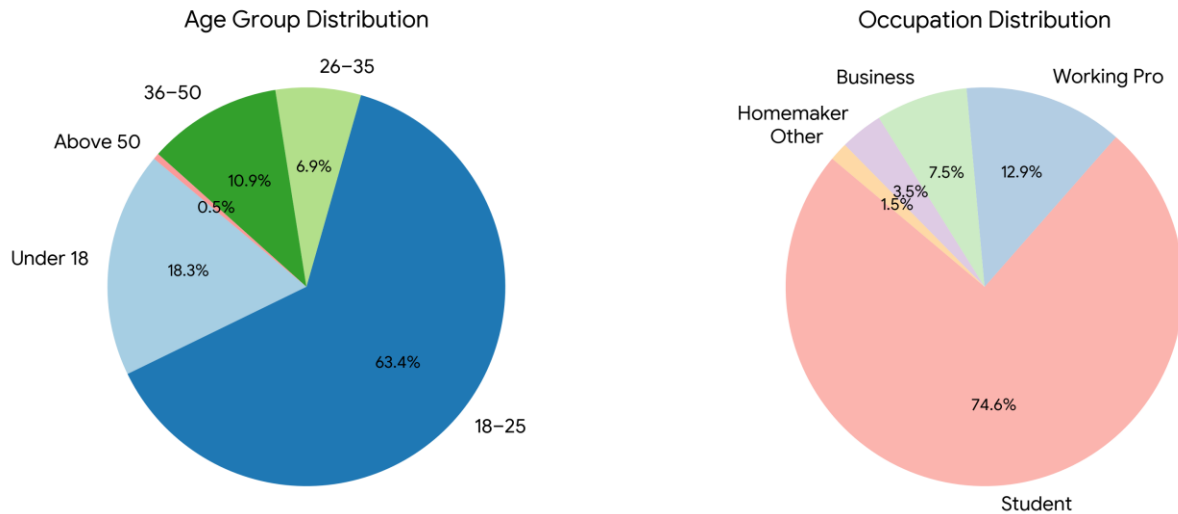
121 Table 1. Demographic Characteristics (N = 202)

Variable	Category	Frequency (N)	Percentage (%)
Age Group	Under 18	37	18.32
	18–25	128	63.37
	26–35	14	6.93
	36–50	22	10.89
	Above 50	1	0.50
Occupation	Student	150	74.26
	Working Professional	26	12.87
	Business Owner	15	7.43

122 **Inference:** *The demographic distribution is heavily concentrated among young adults (63.37% in the 18–25*
 123 *bracket) and students (74.26%). This indicates that the primary adopters of 10-minute delivery services in this Tier-*
 124 *II city are digitally native youth who prioritize convenience and speed in their consumption habits.*

125 As illustrated in Figure 1, the demographic profile displays the distribution of our respondents across age groups.

Figure 1: Demographic Characteristics



126

127 **Reliability and Validity Analysis**

128 Table 2. Reliability and Convergent Validity

Metric	Value	Threshold	Interpretation
Number of Items	7	-	-
Cronbach's Alpha (α)	0.866	> 0.70	Excellent internal consistency
Composite Reliability (CR)	0.903	> 0.70	Highly reliable construct
Average Variance Extracted (AVE)	0.608	> 0.50	Adequate convergent validity

129 **Inference:** Results from the survey instrument are indicative of very strong internal consistency because they
 130 achieve a Cronbach's Alpha of 0.866 and Composite Reliability of 0.903 which are substantially greater than an
 131 acceptable level of 0.70. In addition to these strong internal consistency results, the Average Variance Extracted
 132 (AVE) is also at least adequate as indicated by its value of 0.608 (Hair et al., 2019) and thus provide evidence that
 133 the constructs utilize in the research are reliable in measuring what they are intended to measure.

134 **Descriptive Statistics and Satisfaction Index**

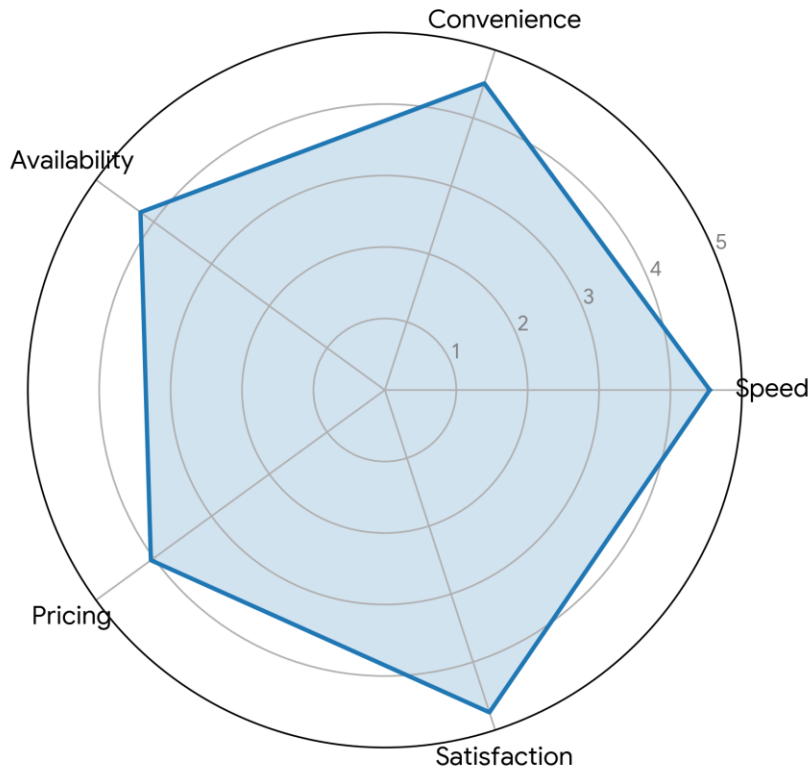
135 Table 3. Item-Level Descriptive Statistics

Survey Item	Mean	SD	Min	Max	Variance
Speed	4.55	0.50	4	5	0.25
Convenience	4.51	0.50	4	5	0.25
Availability	4.23	0.83	3	5	0.70
Pricing	4.05	0.57	3	5	0.33
Overall Satisfaction	4.74	0.44	4	5	0.19

136 **Inference:** Across all service dimensions, respondents provided high ratings with average scores greater than four
 137 point zero (4.0) out of five—for the dimensions rated highest providing average scores in operation at Four Point
 138 Seven Four (4.74) and lowest for pricing at four point zero five (4.05)—which shows that consumers were satisfied
 139 with the operation of the services (e.g., speed & convenience), but were still a little less accepting than expected of
 140 the additional costs associated with those services.

141 As illustrated in Figure 2, the service dimensions (Speed, Convenience, Availability, Pricing, Satisfaction) exhibit a
 142 strong performance profile.

Figure 2: Service Dimensions (Mean Scores)



143

144 Table 4. Satisfaction Index Results

Metric	Result
Mean Satisfaction Score	4.74 / 5.0
Composite Satisfaction Score	958 / 1010
Satisfaction Index (%)	94.85%
Classification	Very High

145 **Inference:** The measured satisfaction index of 94.85% was classified as being within the 'Very High' classification.
 146 As this value is an excessively favourable representation, and communicates an exceptionally high level of market
 147 acceptance and operational effectiveness of q-Commerce platforms for those surveyed in the research study, there
 148 are significant business opportunities for q-commerce solutions in this demographic group.

149 **Correlation Analysis**

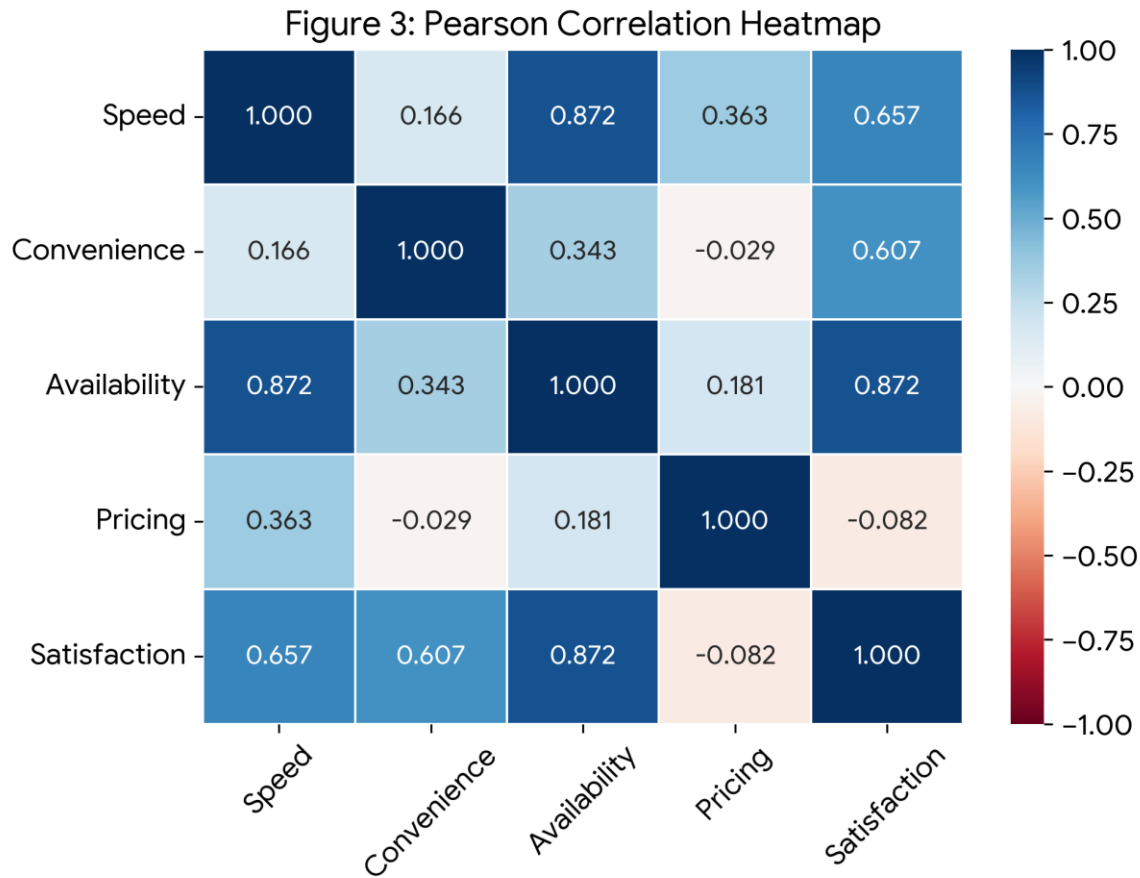
150 Table 5. Pearson Correlation Matrix

Variables	1	2	3	4	5
1. Speed	1.000				
2. Convenience	0.166*	1.000			
3. Availability	0.872***	0.343***	1.000		
4. Pricing	0.363***	-0.029 (ns)	0.181*	1.000	
5. Satisfaction	0.657***	0.607***	0.872***	-0.082 (ns)	1.000

151 Note: *** $p < 0.001$; * $p < 0.05$; ns = not significant.

152 **Inference:** The correlation matrix shows there are high to very high correlations ($r = 0.872$) and ($r = 0.657$) for
 153 availability and speed respectively compared to overall satisfaction; however, pricing ($r = -0.082$) data shows a
 154 weak and statistically insignificant negative relationship to overall satisfaction, indicating that the utility derived
 155 from the platform exceeds any concerns regarding pricing.

156 As illustrated in Figure 3, the correlation heatmap highlights the relationships between Speed, Convenience,
 157 Availability, Pricing, and Satisfaction.



158

159 **Regression Analysis**

160 Table 6. Regression Model Summary

R	R ²	Adjusted R ²	Std. Error
0.956	0.914	0.912	0.130

161 **Inference:** The regression model demonstrates high explanatory power, with an R² value of 0.914 indicating that
 162 91.4% of the variance in overall consumer satisfaction is explained by the independent variables, while the adjusted
 163 R² of 0.912 confirms the robustness of the model.

164 Table 7. ANOVA Results

Model	Sum of Squares	df	Mean Square	F	Sig.
Regression	3.330	4	0.833	521.784	<0.001
Residual	3.330	197	0.017		
Total	38.614	201			

165 **Inference:** The ANOVA results confirm the statistical significance of the regression model ($F = 521.784, p < 0.001$),
 166 validating that the selected variables collectively provide a reliable and significant framework for predicting
 167 consumer satisfaction in the context of this study.

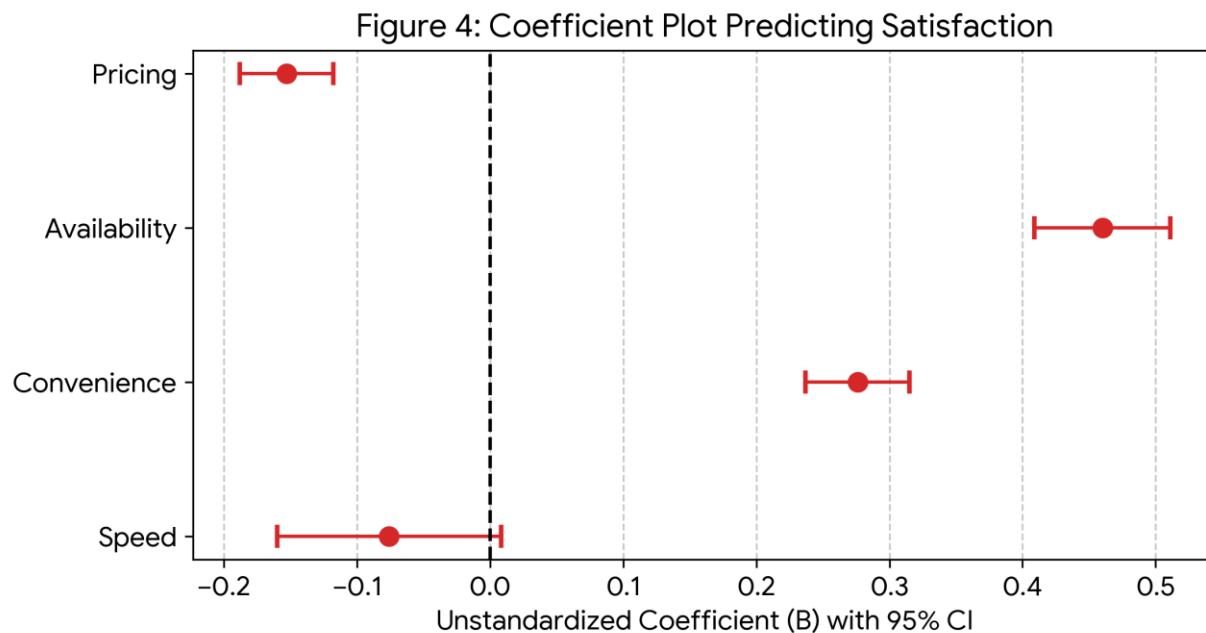
168 Table 8. Regression Coefficients

Predictor	Unstandardized B	Std. Error	Standardized β	t-value	p-value	VIF
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(Constant)	2.518	0.144	-	17.537	<0.001	-
Speed	-0.076	0.043	-0.087	-1.758	0.080	5.54
Convenience	0.276	0.020	0.315	13.545	<0.001	1.24
Availability	0.460	0.026	0.875	17.870	<0.001	5.48
Pricing	-0.153	0.018	-0.200	-8.501	<0.001	1.27

169 **Inference:** Satisfaction has significant positive correlations to Availability ($\beta=0.875$; $p<0.001$) and Convenience
170 ($\beta=0.315$; $p<0.001$) that are strong predictors. Prices also correlate negatively with Satisfaction ($\beta=-0.200$;
171 $p<0.001$). Speed shows a general correlation to Satisfaction; however, after including all variables simultaneously
172 in a multivariate model, the correlation between Speed and Satisfaction becomes marginally non-significant
173 ($p=0.080$) due to the moderate correlation between Speed and Availability ($VIF=5.54$), indicating that when
174 products are readily accessible within one's locality, speed of arrival will not be the decisive baseline metric.

175 As illustrated in Figure 4, the forest plot visualizes the regression coefficients and their confidence intervals.



176

177 Discussion

178 **Evaluation of Service Dimensions and Consumer Satisfaction (H1 to H4)** The multiple linear regression analysis
179 provides a granular breakdown of how individual operational elements dictate overall user satisfaction, yielding
180 distinct conclusions for Hypotheses 1 through 4.

181 • **Product Availability (H3) and Convenience (H2):** There is strong empirical evidence for Hypotheses 3
182 and 2. Product availability is the highest positive driver of customer satisfaction ($\beta=0.875$, $p<0.001$), while
183 platform convenience is also important ($\beta=0.315$, $p<0.001$). As such, a key element of the value proposition
184 of quick commerce is high levels of stock availability and seamless utility.

185 • **Pricing (H4):** Pricing has been found to have a statistically meaningful negative correlation with
186 consumer's overall satisfaction ($\beta=-0.200$), which supports Hypothesis 4. Hence, while there is a
187 considerable degree of utilization of q-commerce by consumers, they still remain conscious about product
188 pricing and additional delivery charges. Nevertheless, the overall high level of satisfaction with the q-
189 commerce service (94.85%) indicates that the value perceived from the service is greater than the
190 consumer's concern with price.

191 • **Delivery Speed (H1):** It is worth noting that the final multivariate model did not support Hypothesis 1.
192 Delivery speed does show a general relationship with satisfaction; however, this becomes marginally non-
193 significant ($p=0.080$) when considered with all other variables in the model. This phenomenon can be

194 attributed to the moderate correlation found between speed and product availability (VIF=5.54). This
195 indicates that when products are available in a local area, the baseline delivery speed is not a strong stand-
196 alone indicator of satisfaction.

197 **Rejection of Demographic Variance (H5 and H6)** The statistical analyses for demographic variables challenge
198 traditional assumptions regarding digital native exclusivity, leading to the rejection of both Hypotheses 5 and 6.

199 • **Age Cohorts (H5):** Statistical insignificance ($p=0.428$, $F=0.964$) from the ANOVA of hypothesis #5
200 (variance between age cohorts) indicates that q-commerce is increasingly being adopted uniformly between
201 age cohorts, as opposed to continuing to be strictly a youth-oriented activity (Eger et al., 2021).

202 • **Occupational Status (H6):** Similarly, Hypothesis 6 was rejected, as statistical testing ($t=-0.463$, $p=0.646$)
203 confirmed that occupational necessity does not dictate usage frequency.

204 **Perception of Functionality over Price** The findings of the regression analysis show that functionality & utility
205 significantly outweigh price in determining customer satisfaction. This shows there is an ever-increasing premium
206 being placed on convenience as part of consumer thinking in India, such that the immediate physical proximity to a
207 retailer—the historic primary benefit of a traditional Kirana store—is becoming increasingly less relevant
208 (Chakraborty et al., 2022).

209 **Hybrid Consumption Patterns** The research also describes overall patterns of change in the way urban consumers
210 interact with their surroundings, with 51% of people saying that they are visiting local Kirana stores less often than
211 they did before. As a result, the traditional retail industry is undergoing fundamental changes across all channels.
212 Abolishing traditional retail will almost certainly not happen, but there will be a mixture of online and offline
213 purchasing methods developing. Consumers are outsourcing their immediate purchases to places like Blinkit, while
214 still continuing to use local stores to make purchase decisions that were made ahead of time (Al Amin et al. 2021).

215 **Practical Implications and Limitations**

216 **Outcomes of The Findings**

217 The results show that traditional retailers can improve their competitive advantage with lower cost digital solutions
218 including WhatsApp-based catalogues to show product offerings and provide visibility into products through digital
219 systems to reduce asymmetric information between retailer and consumer. Additionally, increased development of
220 neighborhood delivery networks will help the local Kirana store deliver the growing consumer demand for
221 convenience and quick fulfilment. For quick commerce companies, a partnership with local Kirana retailers may be
222 a viable operational model. By incorporating local retailers into their platform ecosystems, quick commerce
223 companies will be able to deliver more products to local consumers while improving their last mile delivery
224 capabilities while at the same time, reducing the high cost associated with maintaining independent dark store
225 networks. As such, both traditional retailers and quick commerce companies will have access to mutually beneficial
226 collaboration.

227 **Limitations and Future Research Directions**

228 The study has several limitations in that the research was confined to produce delivered to Bhopal City which may
229 limit generalizability to other areas of India and to other demographic groups. Also, the sample used convenience
230 sampling so that the respondents were predominantly young and digitally connected. As a result, the findings should
231 be interpreted strictly in the context of the particular sample, and not as representative of Indian consumers as a
232 whole. Future research should use probability sampling and multiple city and regional samples to improve the
233 external validity of the sampled population. Longitudinal research could provide insight into how consumer attitudes
234 towards, consumer satisfaction with, and consumer behaviour, will change over time as quick commerce becomes
235 more mainstream. Finally, comparative studies between multiple quick commerce platforms and traditional retailer
236 business models could increase understanding of the changing retail environment.

237

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