



MARKET RESEARCH REPORT

“Analysis of Perishable Product Quality and Brand Trust in Zepto”



BY: PGP GROUP - 10

Anish Dey - MBA25058

Abhishek Joshi - MBA25051

Shashank Kurnal- MBA25226

Rohit Ghosh- MBA25084

Shreya Yadav - MBA25090

Yoganshi - MBA25096

Naman Srivastava - MBATM25014

1. Brief Summary

[Zepto](#), a well-known company in the quick-commerce industry, has recently received more negative reviews that focus on the quality of perishable goods, such as fruits, vegetables, and dairy. This study examines whether particular customer experience variables—spanning from price fairness to perceived reliability—statistically elucidate the variation in "Trust in Freshness."

Using a One-Way ANOVA on a sample of 163 respondents, the study confirms that observed quality, price fairness, and dependability have a significant impact on brand trust. Notably, the "Price Fairness (Q17)" surfaced as the strongest predictor of trust, suggesting that customers heavily equate "value for money" with "freshness".

2. Introduction and Problem Statement

Background:

The quick-commerce model is based on speed and ease of use. But for things that go bad quickly, "freshness" is what makes them valuable. Recent online reviews indicate a discrepancy between what customers expect and what they actually receive when ordering perishables on [Zepto](#).

Problem Statement:

[Zepto](#) is receiving a noticeable share of negative online reviews related to the quality of perishable items. This project aims to identify whether specific customer-experience factors (trust, perceived quality, value fairness, likelihood to repurchase, switching behaviour, etc.) predict or explain these bad reviews.

3. Research Objective

- a) To measure the magnitude of the drop in Brand Trust scores caused by perishable quality failures.
- b) To assess customer satisfaction levels regarding the quality of perishable products delivered by [Zepto](#).
- c) To analyse the factors influencing customers' perceptions of perishable product quality in quick-commerce services.
- d) To evaluate customer response behaviour (churn/switching) when they receive poor-quality perishables and understand its impact on overall platform trust.

4. Research Methodology

- **Research Design:** Descriptive & Causal
- **Data Source:** Primary Data (collected via Google Forms)
- **Sample Size:** N = 150 respondents (Users of [Zepto](#))
- **Scaling:** Likert Scale (1–5) and Binary
- **Analysis Tool:** Microsoft Excel (One-Way ANOVA)
- **Dependent Variable (DV):** Trust in Freshness (Q13)
- **Independent Variables (IV):** Satisfaction (Q14), Reorder Influence (Q15), Reliability (Q16), Price Fairness (Q17), Continuation (Q18), Switching Behaviour (Q19), Recommendation (Q20).

5. Data Analysis and Interpretation

To determine if the Independent Variables (IV) significantly affect the Dependent Variable (Trust in Freshness), One-Way ANOVA tests were conducted. A P-value (<) 0.05 indicates statistical significance.

Summary of ANOVA Results

Dependent Variable: Q13(Trust in Freshness)

Variable	Metric	F-Value	P-Value	Interpretation
Q17	Price Fairness	53.39	1.37E-27	Highest Significance
Q15	Reorder Influence	52.05	6.10E-23	Highly Significant
Q14	Satisfaction w/ Quality	35.32	1.40E-20	Highly Significant
Q16	Reliable Quality Controls	28.95	1.08E-17	Highly Significant
Q19	Switching Behaviour	26.04	2.81E-16	Highly Significant
Q20	Recommendation Likelihood	4.50	0.0018	Significant
Q18	Continuation (Loyalty)	2.44	0.049	Marginally Significant

Detailed Interpretation by Factor

Factor 1: Price Fairness (Q17)

- **Result:** $F = 53.39, P < 0.001$
- **Analysis:** This variable had the highest F-value, showing the strongest relationship with trust.
- **Insight:** Customers who felt the pricing was unfair correlated with low trust in freshness. This suggests that when customers receive poor-quality produce, they feel cheated on price, resulting in a lowering of their trust.

Factor 2: Satisfaction with Quality (Q14) & Reliability (Q16)

- **Result (Q14):** $F = 35.32, P < 0.001$
- **Result (Q16):** $F = 28.95, P < 0.001$
- **Analysis:** Both variables are highly significant.
- **Insight:** There is a direct causal link between current satisfaction/reliability and the deeper metric of "Trust." A failure in daily quality control (Q16) directly results in degradation of the long-term trust score (Q13).

Factor 3: Switching Behaviour (Q19)

- **Result:** $F = 26.04, P < 0.001$
- **Analysis:** Highly significant.
- **Insight:** This addresses Objective 4. Customers with low trust in freshness are likely to switch to the existing competitors. The variance in trust explains the variance in churn intent.

Factor 4: Continuation/Loyalty (Q18)

- **Result:** $F = 2.44, P = 0.049$
- **Analysis:** This is the weakest link among the factors. Although it is statistically significant (with a p-value < 0.05), the relationship is not strong.
- **Insight:** This suggests a "Convenience Trap." Customers may continue using [Zepto](#) (Q18) despite not trusting its freshness (Q13), as they rely on its speed and convenience. However, this loyalty will become fragile over time.

6. Findings

Based on the statistical analysis, we get to the following finding that directly mapped to the research objectives:

- 1. Trust is Fragile (Objective 1):** When quality is compromised, trust declines significantly. The extremely low P-values for all quality metrics (Q14, Q16) indicate that trust is weak and quickly erodes when quality issues arise.
- 2. Price Fairness is the Anchor (Objective 3):** The most significant factor affecting perception is not just the visual quality of the fruit and vegetables, but also Value Fairness. If a customer pays extra for fast delivery and receives subpar avocados, the loss of trust becomes even greater due to the financial impact.
- 3. Churn is Imminent (Objective 4):** There is a clear link between Trust in Freshness (Q13) and Switching Behaviour (Q19). Poor quality of perishable items strongly predicts customer churn to competitors like Blinkit or Instamart.
- 4. Sad Reviews Proxy (Q20):** The likelihood to recommend (Q20) is significant ($P=0.0018\$$) but not as much as Price or Reliability. This suggests that customers might not immediately leave negative reviews, but they will lose trust and stop buying.

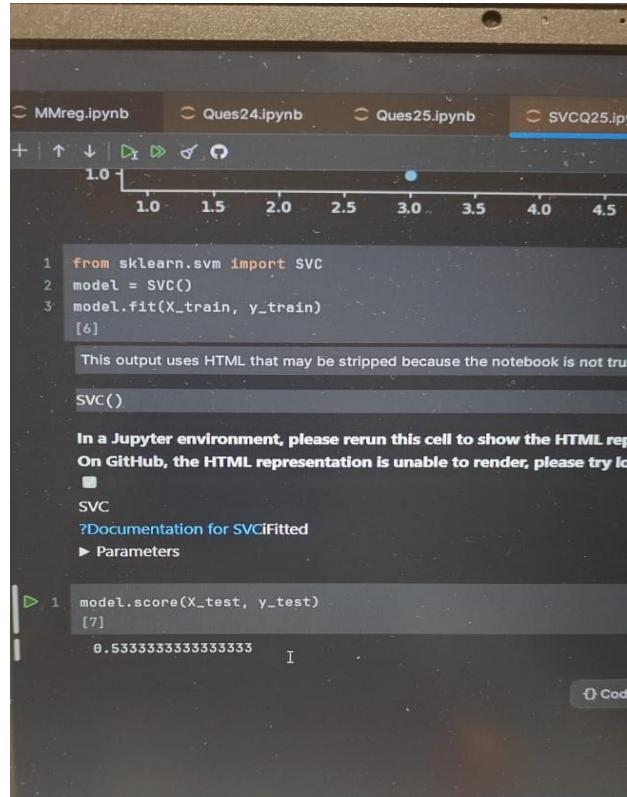
7. Recommendations

- 1. Set up "Fairness" Refunds:** Q17 (Price Fairness) is the most critical factor in building trust, so [Zepto](#) needs to have a "No-Questions-Asked" instant refund policy for perishable goods. If the customer thinks the quality wasn't worth the price, giving them their money back immediately can help maintain their trust score.
- 2. Quality Control at Dark Stores:** The significance of Q16 (Reliability) suggests the issue lies in the variability. [Zepto](#) needs stricter grading protocols at the dark store level before packing.
- 3. Separate "Fresh" Feedback Loop:** Create a specific post-order feedback flow for fruits/veggies. Since Q18 (Continuation) is weak, [Zepto](#) cannot rely on general retention metrics to assume customers are happy with freshness. They are staying for speed, but they are buying perishables elsewhere.

8. Conclusion

The Null Hypothesis (H_0) is rejected for all variables. Customer perceptions of quality, price fairness, and reliability significantly affect their trust in [Zepto's](#) ability to deliver fresh items. The data suggests that while customers may seek speed and stay for it (weak correlation in Q18), they are also considering switching (strong correlation in Q19) if the "Value for Money" regarding the freshness of perishable goods is not rectified and justified.

ANNEXURE



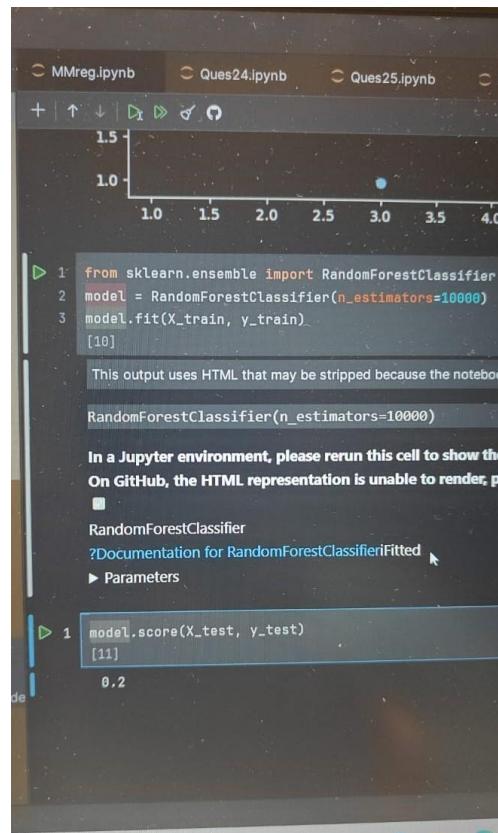
A screenshot of a Jupyter Notebook interface. The top navigation bar shows tabs for 'MMreg.ipynb', 'Ques24.ipynb', 'Ques25.ipynb', and 'SVCQ25.ipynb' (the active tab). Below the tabs is a toolbar with icons for cell operations. A plot window displays a single point at (3.0, 1.0) on a coordinate system with axes from 1.0 to 4.5. The main code cell contains the following Python code:

```
1 from sklearn.svm import SVC
2 model = SVC()
3 model.fit(X_train, y_train)
[6]
```

Below the code, a message states: "This output uses HTML that may be stripped because the notebook is not trusted". The output cell shows the result of the fit command: "SVC()". A tooltip for 'SVC' provides documentation and parameters information. Another code cell below shows the score method:

```
1 model.score(X_test, y_test)
[7]
0.5333333333333333
```

SUPPORT VECTOR CLASSIFICATION

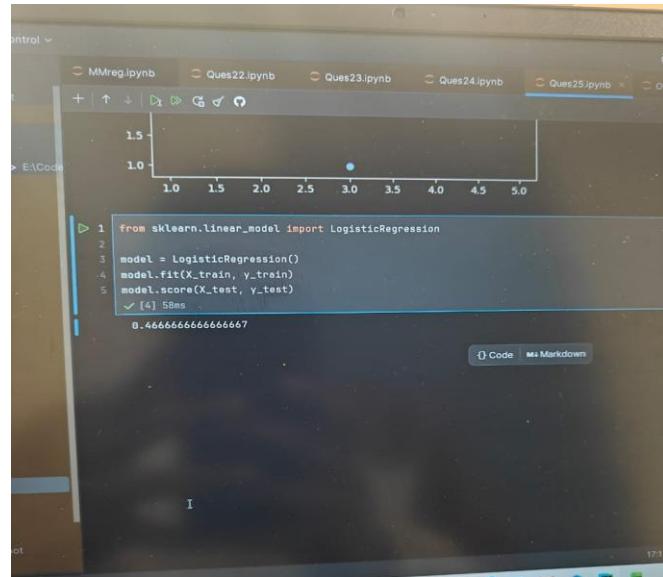


A screenshot of a Jupyter Notebook interface. The top navigation bar shows tabs for 'MMreg.ipynb', 'Ques24.ipynb', 'Ques25.ipynb', and 'SVCQ25.ipynb'. Below the tabs is a toolbar with icons for cell operations. A plot window displays a single point at (3.0, 1.0) on a coordinate system with axes from 1.0 to 4.0. The main code cell contains the following Python code:

```
1 from sklearn.ensemble import RandomForestClassifier
2 model = RandomForestClassifier(n_estimators=1000)
3 model.fit(X_train, y_train)
[10]
```

Below the code, a message states: "This output uses HTML that may be stripped because the notebook is not trusted". The output cell shows the result of the fit command: "RandomForestClassifier(n_estimators=1000)". A tooltip for 'RandomForestClassifier' provides documentation and parameters information. Another code cell below shows the score method:

```
1 model.score(X_test, y_test)
[11]
0.2
```



LOGISTIC REGRESSION

A screenshot of a Jupyter Notebook interface. The top navigation bar shows several open notebooks: Ques24.ipynb, KFold.ipynb (which is the active tab), Ques25.ipynb, and SVCQ25.ipynb. The current notebook tab is KFold.ipynb. Below the tabs is a toolbar with icons for back, forward, search, and other notebook functions. The main area contains code for KFold cross-validation:

```
1 %matplotlib inline
2 df = pd.read_excel(r"D:\Data Science\Responses.xlsx")
3
4 from sklearn.model_selection import StratifiedKFold
5 folds = StratifiedKFold(n_splits=10, shuffle=True)
6
7 from sklearn.model_selection import cross_val_score
8 from sklearn.model_selection import train_test_split
9
10 X_train, X_test, y_train, y_test = train_test_split(df[['Q25_num']],
11                                                 df.Q13_Trust_Zepeto,
12                                                 test_size=0.2, random_state=42)
13
14 cross_val_score(LogisticRegression(), X_train, y_train, cv=2)
15
16 array([0.48529412, 0.49253731])
17
18 cross_val_score(SVC(), X_train, y_train, cv=2)
19
20 array([0.41176471, 0.37313433])
21
22 cross_val_score(RandomForestClassifier(), X_train, y_train, cv=2)
23
24 array([0.41176471, 0.35820896])
```

The code cell has a status bar indicating it took 225ms to run. At the bottom right of the notebook area, there are buttons for "Code" and "Markdown".

KFold Cross Validation