

Cash Flow vs. Loyalty Assessment (Finance)

Language: PostgreSQL

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1  /*
2  BUSINESS QUESTION:
3  "If we restrict high-installment credit plans, will it hurt our
4  Champions or mainly impact low-value segments?"
5
6  WHY THIS MATTERS:
7  High-installment plans (8-12 months) are expensive for the business
8  (interchange fees, credit risk, cash flow delay). But if Champions
9  depend on them, restricting installments could backfire. This query
10 breaks down credit dependency by segment so Finance can model the
11 impact before changing payment policy.
12 */
13
14 WITH delivered_base AS (
15     SELECT
16         c.customer_unique_id,
17         o.order_id,
18         o.order_purchase_timestamp
19     FROM olist_customers_dataset AS c
20     INNER JOIN olist_orders_dataset AS o
21         ON o.customer_id = c.customer_id
22     WHERE o.order_status = 'delivered'
23 ),
24
25 rfm_metrics AS (
26     SELECT
27         db.customer_unique_id,
28         (SELECT MAX(order_purchase_timestamp)::DATE FROM delivered_base)
29         - MAX(db.order_purchase_timestamp)::DATE AS recency_days,
30         COUNT(DISTINCT db.order_id) AS frequency,
31         ROUND(SUM(p.payment_value)::NUMERIC, 2) AS monetary
32     FROM delivered_base AS db
33     INNER JOIN olist_order_payments_dataset AS p
34         ON p.order_id = db.order_id
35     GROUP BY db.customer_unique_id
36 ),
37
38 rfm_scores AS (
39     SELECT *,
40         NTILE(5) OVER (ORDER BY recency_days DESC) AS r_score,
41         NTILE(5) OVER (ORDER BY frequency ASC) AS f_score,
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42         NTILE(5) OVER (ORDER BY monetary      ASC)      AS m_score
43     FROM rfm_metrics
44 ),
45
46 rfm_segments AS (
47     SELECT *,
48         CASE
49             WHEN r_score >= 4 AND f_score >= 4 AND m_score >= 4 THEN 'Champions'
50             WHEN r_score <= 2 AND f_score >= 4 AND m_score >= 4 THEN 'At Risk'
51             WHEN r_score >= 4 AND f_score <= 2 AND m_score <= 2 THEN 'New Customers'
52             WHEN r_score >= 4 AND (f_score >= 3 OR m_score >= 3) THEN 'Potential Loyalists'
53             WHEN r_score >= 3 AND f_score >= 3 AND m_score >= 3 THEN 'Loyal Customers'
54             WHEN r_score <= 2 AND f_score <= 2 AND m_score <= 2 THEN 'Hibernating'
55             ELSE 'Need Attention'
56         END AS customer_segment
57     FROM rfm_scores
58 ),
59
60 /* ■■ Financial dimension: dominant payment type + avg installments ■■ */
61 dim_financial AS (
62     SELECT
63         f_type.customer_unique_id,
64         f_type.dominant_payment_type,
65         COALESCE(f_inst.avg_installments, 0)           AS avg_installments,
66         COALESCE(f_inst.max_installments, 0)          AS max_installments
67     FROM (
68         SELECT customer_unique_id, payment_type AS dominant_payment_type
69         FROM (
70             SELECT db.customer_unique_id, p.payment_type,
71                 ROW_NUMBER() OVER (PARTITION BY db.customer_unique_id
72                                     ORDER BY COUNT(*) DESC, SUM(p.payment_value) DESC) AS rn
73             FROM delivered_base           AS db
74             INNER JOIN olist_order_payments_dataset AS p ON p.order_id = db.order_id
75             GROUP BY db.customer_unique_id, p.payment_type
76         ) ranked WHERE rn = 1
77     ) f_type
78     LEFT JOIN (
79         SELECT db.customer_unique_id,
80             ROUND(AVG(p.payment_installments):NUMERIC, 1) AS avg_installments,
81             MAX(p.payment_installments)                    AS max_installments
82         FROM delivered_base           AS db
83         INNER JOIN olist_order_payments_dataset AS p ON p.order_id = db.order_id
84         WHERE p.payment_type = 'credit_card' AND p.payment_installments > 0
85         GROUP BY db.customer_unique_id
86     ) f_inst ON f_inst.customer_unique_id = f_type.customer_unique_id
87 ),
88

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89  /* ■■ Revenue by payment type per customer (for boleto % calc) ■■ */
90  payment_revenue AS (
91      SELECT
92          db.customer_unique_id,
93          SUM(p.payment_value)                AS total_paid,
94          SUM(CASE WHEN p.payment_type = 'boleto'
95              THEN p.payment_value ELSE 0 END) AS boleto_paid,
96          SUM(CASE WHEN p.payment_type = 'credit_card'
97              THEN p.payment_value ELSE 0 END) AS credit_card_paid
98      FROM delivered_base                    AS db
99      INNER JOIN olist_order_payments_dataset AS p ON p.order_id = db.order_id
100     GROUP BY db.customer_unique_id
101 )
102
103  /* ■■ FINAL: Credit dependency profile by segment ■■ */
104  SELECT
105     rfm.customer_segment,
106     COUNT(*)                AS total_customers,
107     ROUND(SUM(rfm.monetary)::NUMERIC, 2) AS total_segment_revenue,
108
109     /* Average installments used (credit card only) */
110     ROUND(AVG(fin.avg_installments)::NUMERIC, 1) AS avg_cc_installments,
111
112     /* % of customers in this segment who max out at 6+ installments -
113        these are the "installment-dependent" buyers */
114     ROUND(100.0 *
115         SUM(CASE WHEN fin.max_installments >= 6 THEN 1 ELSE 0 END)::NUMERIC
116         / COUNT(*)
117     , 1) AS pct_using_6plus_installments,
118
119     /* Revenue concentration by payment method */
120     ROUND(100.0 * SUM(pr.boleto_paid)::NUMERIC
121         / NULLIF(SUM(pr.total_paid), 0), 1) AS pct_revenue_via_boleto,
122     ROUND(100.0 * SUM(pr.credit_card_paid)::NUMERIC
123         / NULLIF(SUM(pr.total_paid), 0), 1) AS pct_revenue_via_credit_card,
124
125     /* Dominant payment type distribution */
126     SUM(CASE WHEN fin.dominant_payment_type = 'credit_card' THEN 1 ELSE 0 END)
127     AS count_dominant_cc,
128     SUM(CASE WHEN fin.dominant_payment_type = 'boleto' THEN 1 ELSE 0 END)
129     AS count_dominant_boleto
130
131  FROM rfm_segments AS rfm
132  LEFT JOIN dim_financial AS fin ON fin.customer_unique_id = rfm.customer_unique_id
133  LEFT JOIN payment_revenue AS pr ON pr.customer_unique_id = rfm.customer_unique_id
134
135  GROUP BY rfm.customer_segment

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