



GCash:

User Engagement Analysis

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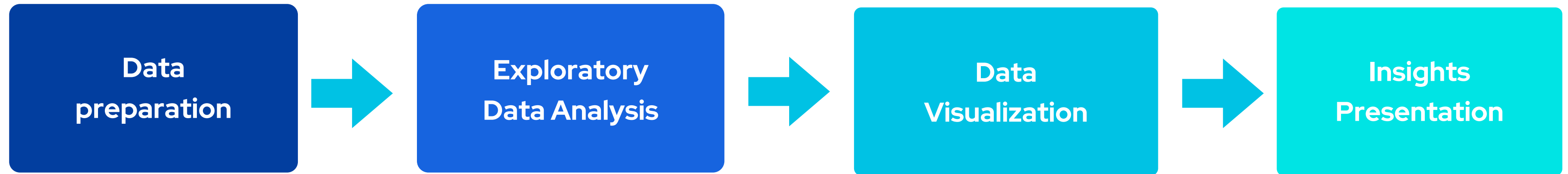
Objective

Understand user engagement patterns to identify key opportunities for improving user experience and retention.

Scope

- Analyze app usage and transaction behaviors
- Segment users based on engagement
- Identify trends by location, transaction type, and merchant
- Perform cohort analysis

Steps taken for analysis



Process Flow

Data Preparation

Datasets Used

Transactions
(raw data)

transactions						
user_id	transaction_id	transaction_date	transaction_amount	transaction_type	merchant_id	
645943	1	5/14/2024	176	scan to pay	39	
645240	2	2/13/2024	154	bills pay	8	
645787	3	4/6/2024	375	buy load	16	
645512	4	1/28/2024	298	gsave	1	
646989	5	1/15/2024	372	web pay	34	
645075	6	3/13/2024	582	bank transfer	3	
645553	7	5/17/2024	313	g life	25	
646960	8	4/27/2024	103	scan to pay	36	
645806	9	5/20/2024	340	bills pay	15	
646918	10	2/21/2024	923	buy load	18	

Data
Preparation

Datasets Used

Users

users				
user_id	registration_date	age	#	location
645000	11/9/2023	39	NL	
645001	3/6/2023	15	NL	
645002	11/6/2023	30	SL	
645003	11/26/2023	25	GMA+	
645004	12/1/2023	36	GMA+	
645005	5/5/2023	44	GMA+	
645006	4/18/2023	50	GMA+	
645007	5/30/2023	31	GMA	
645008	5/15/2023	34	SL	
645009	12/24/2023	44	GMA+	

1. Remove Duplicates

```
3  -- REMOVE DUPLICATE ROWS FROM RAW DATA TABLE
4  • SELECT
5      *
6  FROM gcash.transactions_staging;
7
8  • WITH duplicate_cte AS ( -- to see if there's any duplicate row
9      SELECT *,
10         ROW_NUMBER() OVER (
11             PARTITION BY user_id, transaction_id, transaction_date, transaction_amount, transaction_type, merchant_id
12             ORDER BY transaction_id
13         ) AS row_num
14      FROM gcash.transactions_staging
15  )
16  SELECT *
17  FROM duplicate_cte
18  WHERE row_num > 1;
```

Query result:

user_id	transaction_id	transaction_date	transaction_amount	transaction_type	merchant_id	row_num
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2. Standardized Formats

```
-- Data standardization
UPDATE gcash.transactions_staging
-- to convert all transaction_type to lowercase and remove extra spaces.
SET
    transaction_type = LOWER(TRIM(transaction_type));
```

2. Standardized Formats

```
-- convert the data type of registration_date under users table from text to date type
SELECT -- to ensure that all dates in the column have the same format
    u.registration_date,
    STR_TO_DATE(u.registration_date, '%m/%d/%Y')
FROM
    gcash.users u;

UPDATE -- actual change the format from its raw form
    gcash.users
SET
    users.registration_date = STR_TO_DATE(registration_date, '%m/%d/%Y')
WHERE
    users.registration_date IS NOT NULL
    AND users.registration_date <> '';

ALTER TABLE gcash.users
MODIFY COLUMN registration_date DATE; -- to change data type from text to date
```

3. Check for null values under transactions table

```
SELECT -- to check for any null values for each columns in raw data table
SUM(CASE WHEN user_id IS NULL THEN 1 ELSE 0 END) AS null_user_id,
SUM(CASE WHEN transaction_id IS NULL THEN 1 ELSE 0 END) AS null_transaction_id,
SUM(CASE WHEN transaction_date IS NULL THEN 1 ELSE 0 END) AS null_transaction_date,
SUM(CASE WHEN transaction_amount IS NULL THEN 1 ELSE 0 END) AS null_transaction_amount,
SUM(CASE WHEN transaction_type IS NULL OR transaction_type = '' THEN 1 ELSE 0 END) AS null_transaction_type,
SUM(CASE WHEN merchant_id IS NULL THEN 1 ELSE 0 END) AS null_merchant_id
FROM gcash.transactions_staging;
```

Query result:

	null_user_id	null_transaction_id	null_transaction_date	null_transaction_amount	null_transaction_type	null_merchant_id
▶	0	0	0	0	0	0

4. Join transactions table to users table for data visualization

```
-- now let's join raw data table to users table for our data analysis and data visualization
SELECT -- to combine the raw data table to users table which will be used for data visualization
    t.user_id,
    t.transaction_id,
    t.transaction_date,
    t.transaction_amount,
    t.transaction_type,
    t.merchant_id,
    u.registration_date,
    u.age,
    CASE -- to change from integers to the actual location name using the reference table
        WHEN u.location = 1 THEN 'VisMin'
        WHEN u.location = 2 THEN 'SL'
        WHEN u.location = 3 THEN 'NL'
        WHEN u.location = 4 THEN 'GMA+'
        WHEN u.location = 5 THEN 'GMA'
        ELSE 'Unknown'
    END AS location
FROM gcash.transactions_staging t
JOIN gcash.users1 u
ON t.user_id = u.user_id;
```

4. Join transactions table to users table for data visualization

Query result:

	user_id	transaction_id	transaction_date	transaction_amount	transaction_type	merchant_id	registration_date	age	location
▶	645943	1	2024-05-14	176	scan to pay	39	8/4/2023	27	GMA
	645240	2	2024-02-13	154	bills pay	8	2/19/2023	32	NL
	645787	3	2024-04-06	375	buy load	16	11/3/2023	23	SL
	645512	4	2024-01-28	298	gsave	1	4/29/2023	50	VisMin
	646989	5	2024-01-15	372	web pay	34	4/25/2023	38	NL
	645075	6	2024-03-13	582	bank transfer	3	7/18/2023	29	GMA+
	645553	7	2024-05-17	313	g life	25	5/6/2023	32	VisMin
	646960	8	2024-04-27	103	scan to pay	36	1/24/2023	30	GMA
	645806	9	2024-05-20	340	bills pay	15	5/17/2023	23	NL

Exploratory Data Analysis

Initial Calculations

```
SELECT -- total transactions per region
  u.location,
  COUNT(t.user_id) AS total_transactions_per_location
FROM gcash.transactions_staging t
  JOIN gcash.users u
  ON t.user_id = u.user_id
GROUP BY
  1;
```

	location	total_transactions_per_location
▶	GMA	929
	NL	999
	SL	1106
	VisMin	998
	GMA+	968

```
SELECT -- top transaction type
  t.transaction_type,
  COUNT(t.transaction_type) AS most_used_transactions
FROM
  gcash.transactions_staging t
GROUP BY
  1
ORDER BY
  2 DESC;
```

	transaction_type	most_used_transactions
▶	gsave	999
	web pay	993
	bank transfer	751
	g life	751
	scan to pay	502
	bills pay	502
	buy load	502

Initial Calculations

```
SELECT -- leading merchants
       t.merchant_id,
       COUNT(t.merchant_id) AS most_used_merchants
FROM   gcash.transactions_staging t
GROUP BY
       1
ORDER BY
       2 DESC;
```

	merchant_id	most_used_merchants
▶	4	422
	2	421
	1	403
	3	371
	21	214
	25	210
	23	204
	24	193
	22	180
	5	133
	17	114

```
SELECT -- monthly transaction volume
       DATE_FORMAT(t.transaction_date, '%Y-%m') AS month,
       AVG(t.transaction_amount) AS avg_transaction_amount,
       SUM(t.transaction_amount) AS total_transaction_amount
FROM   gcash.transactions_staging t
GROUP BY
       1
ORDER BY
       1;
```

	month	avg_transaction_amount	total_transaction_amount
▶	2024-01	550.0555	605061
	2024-02	547.7566	560355
	2024-03	541.6654	574707
	2024-04	534.6933	568379
	2024-05	528.8008	398187

Cohort Analysis

```
SELECT -- cohort analysis for user retention
    DATE_FORMAT(u.registration_date, '%Y-%m') AS cohort_month,
    TIMESTAMPDIFF(MONTH, u.registration_date, t.transaction_date) AS months_since_signup,
    COUNT(DISTINCT t.user_id) AS active_users
FROM
    gcash.transactions_staging t
    JOIN gcash.users u
    ON t.user_id = u.user_id
WHERE
    t.transaction_date >= u.registration_date
GROUP BY
    cohort_month, months_since_signup
ORDER BY
    cohort_month, months_since_signup;
```

Cohort Analysis

Query result:

	cohort_month	months_since_signup	active_users
►	2023-01	11	36
	2023-01	12	72
	2023-01	13	65
	2023-01	14	76
	2023-01	15	63
	2023-01	16	19
	2023-02	10	25
	2023-02	11	69
	2023-02	12	53
	2023-02	13	69
	2023-02	14	74
	2023-02	15	21
	2023-03	9	44
	2023-03	10	74
	2023-03	11	57
	2023-03	12	64
	2023-03	13	59
	2023-03	14	16
	2023-04	8	31
	2023-04	9	66
	2023-04	10	56

User Segmentation

```
SELECT
    AVG(t.transaction_amount) AS avg_transaction_amount,
    MIN(t.transaction_amount) AS min_transaction_amount,
    MAX(t.transaction_amount) AS max_transaction_amount
FROM
    gcash.transactions_staging t;
```

Query result:

	avg_transaction_amount	min_transaction_amount	max_transaction_amount
►	541.3378	100	1000

```
SELECT
    AVG(user_transaction_count) AS avg_transaction_count,
    MAX(user_transaction_count) AS max_transaction_count,
    MIN(user_transaction_count) AS min_transaction_count
FROM
    (SELECT
        t.user_id,
        COUNT(user_id) AS user_transaction_count
    FROM
        gcash.transactions_staging t
    GROUP BY
        1
    ) as txn_per_user;
```

Query result:

	avg_transaction_count	max_transaction_count	min_transaction_count
►	2.7397	11	1

User Segmentation

Frequency Score (1-3)

3 points =
Transaction Count ≥ 5

2 points =
Transaction Count ≥ 2

1 point =
Transaction Count < 2

Value Score (1-3)

3 points =
Transaction Amount ≥ 700

2 points =
Transaction Amount ≥ 400

1 point =
Transaction Amount < 400

Recency Score (1)

1 point = Any transaction
within April 22, 2024 to May
21, 2024

User Segmentation

1 - 3

=

Low Engagement

4 - 6

=

At Risk

7

=

Highly Engaged

User Segmentation

```
-- Now that we already computed for the average, maximum,
-- and minimum amount of both transaction count and transaction
-- amount, let's move on to user segmentation through engagement score
) WITH user_stats AS (
    SELECT
        t.user_id,
        COUNT(*) AS txn_count,
        SUM(t.transaction_amount) AS total_txn_value,
        AVG(t.transaction_amount) AS avg_txn_value,
        MAX(t.transaction_date) AS last_txn_date,

        -- Frequency score
        CASE
            WHEN COUNT(*) >= 5 THEN 3
            WHEN COUNT(*) >= 2 THEN 2
            ELSE 1
        END AS freq_score,

        -- Value score
        CASE
            WHEN AVG(t.transaction_amount) >= 700 THEN 3
            WHEN AVG(t.transaction_amount) >= 400 THEN 2
            ELSE 1
        END AS value_score,

        -- Recency score
        CASE
            WHEN MAX(t.transaction_date) >= DATE_SUB('2024-05-21', INTERVAL 30 DAY) THEN 1
            -- this is to see if there's a transaction made within the last 30 days of the available data
            ELSE 0
        END AS recency_score

    FROM gcash.transactions_staging t
    GROUP BY t.user_id
```

User Segmentation

```
SELECT
  user_id,
  txn_count,
  total_txn_value,
  avg_txn_value,
  last_txn_date,
  freq_score,
  value_score,
  recency_score,
  (freq_score + value_score + recency_score) AS total_score,

  -- Segment based on score
  CASE
    WHEN (freq_score + value_score + recency_score) = 7 THEN 'Highly Engaged'
    WHEN (freq_score + value_score + recency_score) BETWEEN 4 AND 6 THEN 'At Risk'
    ELSE 'Low Engagement'
  END AS engagement_segment

FROM
  user_stats

ORDER BY
  total_score;
```

	user_id	txn_count	total_txn_value	avg_txn_value	last_txn_date	freq_score	value_score	recency_score	total_score	engagement_segment
▶	646396	1	306	306.0000	2024-02-15	1	1	0	2	Low Engagement
	646387	1	219	219.0000	2024-04-01	1	1	0	2	Low Engagement
	645350	1	316	316.0000	2024-03-03	1	1	0	2	Low Engagement
	646469	1	193	193.0000	2024-02-03	1	1	0	2	Low Engagement
	646763	1	252	252.0000	2024-03-29	1	1	0	2	Low Engagement
	645346	1	379	379.0000	2024-03-30	1	1	0	2	Low Engagement
	646586	1	129	129.0000	2024-03-24	1	1	0	2	Low Engagement
	645700	1	291	291.0000	2024-02-22	1	1	0	2	Low Engagement
	646268	1	157	157.0000	2024-01-02	1	1	0	2	Low Engagement
	645907	1	104	104.0000	2024-03-01	1	1	0	2	Low Engagement
	646110	1	174	174.0000	2024-02-08	1	1	0	2	Low Engagement

Data Visualization

1825

Total Active Users

5000

Total Transaction Count

₱2.71M

Total Transaction Amount

₱541.34

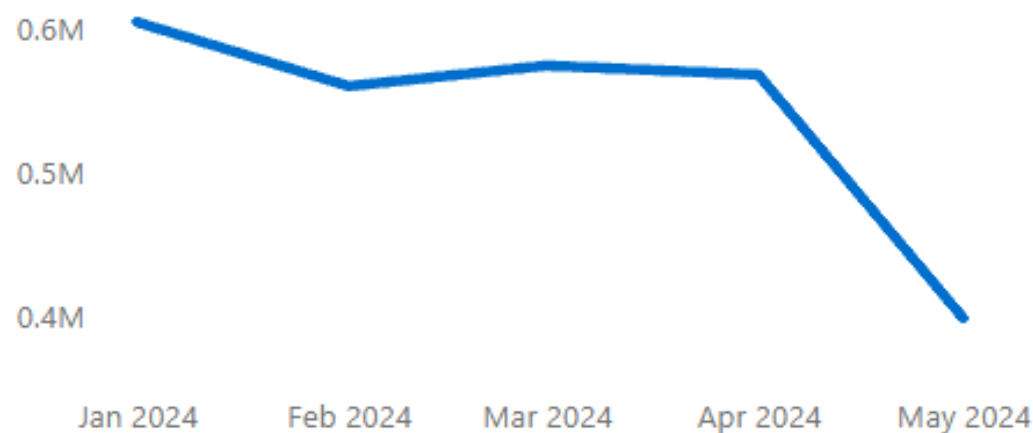
Average Transaction Amount

Date
1/1/2024 5/21/2024

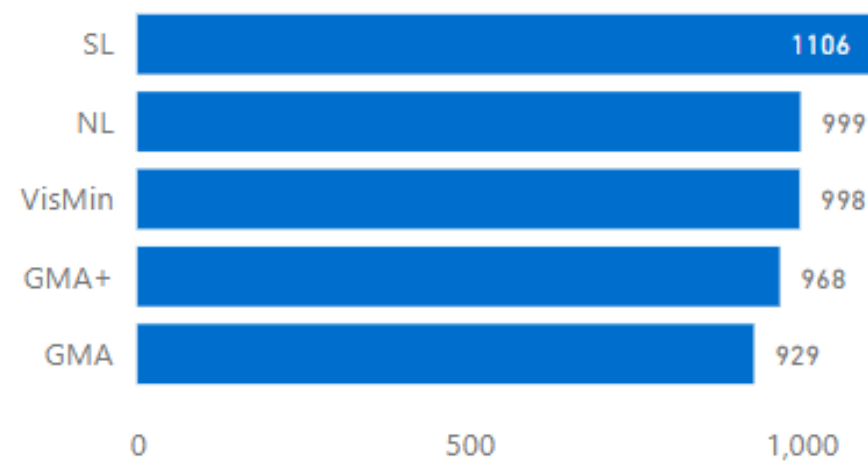
Location
All

Cohort Month
All

Transaction Amount Over Time



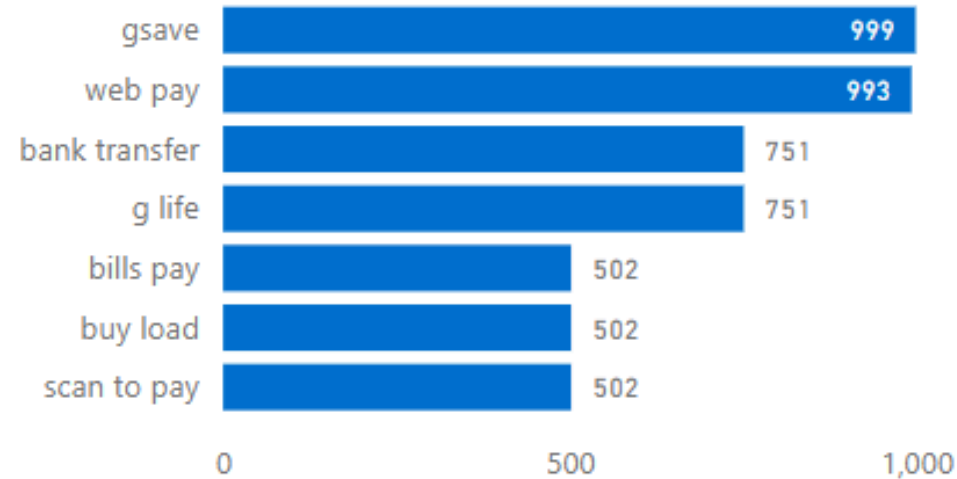
Transaction Count by Location



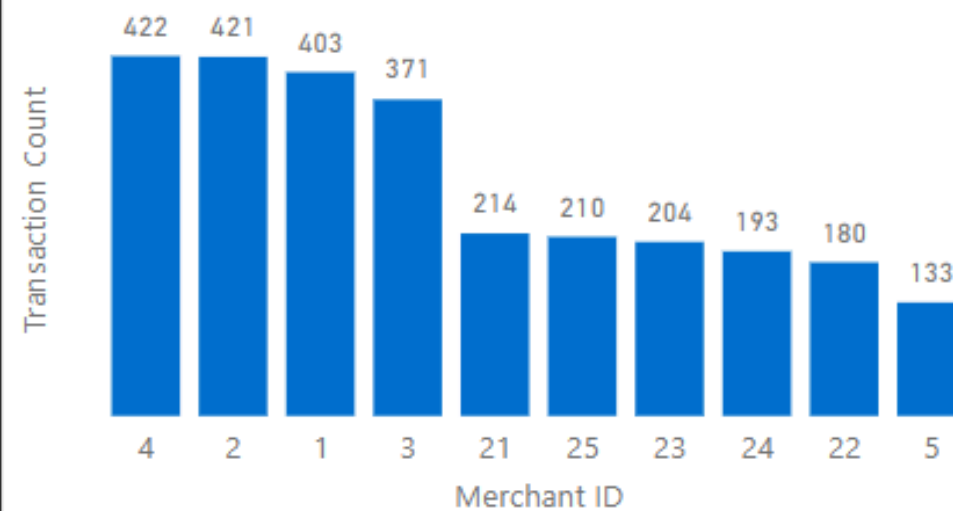
User Cohort Analysis

Year	Month	months_since_signup	active_users
2023	January	16	19
2023	January	11	36
2023	January	15	63
2023	January	13	65
2023	January	12	72

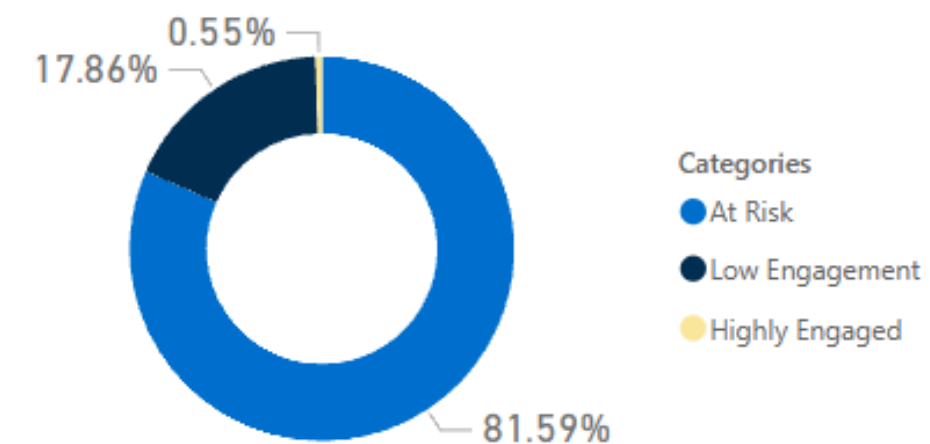
Top Transaction Types



Top 10 Merchant IDs



User Engagement Segmentation



Insights

Cohort Analysis

User Segmentation

Early Drop-
off Is
Consistent
Across
Cohorts

Short-
term
engage
ment
focus

Majority are
'At Risk'
segment and
minimal
highly
engaged
users

Declining transaction
trend signals
retention concerns.

Location-based gaps
point to uneven app
penetration.

Engagement is
anchored on financial
transactions.

E-commerce players
show high transaction
frequency.

Online gaming and
betting are emerging
engagement drivers.

Thank you