- -- Flipkart Data Analysis using SQL
- -- Solutions of 20 business problems

1 Understand how many unique users are active on the platform each day to track daily engagement trends (DAU – Daily Active Users).

SELECT activity\_date, COUNT(DISTINCT user\_id) AS daily\_active\_users
FROM user\_activity
GROUP BY activity\_date
ORDER BY activity\_date;

2 Identify how many distinct users use the platform each month to analyze user base growth and retention (MAU – Monthly Active Users).

SELECT DATE\_TRUNC('month', activity\_date) AS month, COUNT(DISTINCT user\_id) AS monthly\_active\_users FROM user\_activity GROUP BY month ORDER BY month;

3 Monitor how many new users are signing up on the platform to evaluate acquisition performance (New Signups).

SELECT activity\_date, COUNT(DISTINCT user\_id) AS new\_signups FROM user\_activity
WHERE is\_new\_user = TRUE
GROUP BY activity\_date
ORDER BY activity\_date;

4. Measure the month-over-month growth or decline in the user base to adjust growth strategies effectively (MoM – Month-over-Month Growth).

5. Analyze which traffic sources (e.g., Organic, Ads, Referral) bring the most users to optimize marketing efforts (Traffic Source Analysis).

SELECT source, COUNT(DISTINCT user\_id) AS users FROM user\_activity
GROUP BY source
ORDER BY users DESC;

6. Understand whether users prefer using the app or the web version to improve UX and prioritize platform-specific features (Device Preference – App vs Web).

SELECT platform, COUNT(DISTINCT user\_id) AS users FROM user\_activity GROUP BY platform;

7.Identify how many users are visiting for the first time versus returning users to understand retention and loyalty (New vs Returning Users).

SELECT is\_new\_user, COUNT(DISTINCT user\_id) AS users FROM user\_activity

8. Determine how many users signed up through referrals to assess the effectiveness of the referral program (Referral Signups).

SELECT COUNT(DISTINCT user\_id) AS referral\_signups FROM user\_activity WHERE referral user id IS NOT NULL;

9.Analyze the user drop-off rate at each stage of the funnel (View  $\rightarrow$  Cart  $\rightarrow$  Purchase) to optimize conversion (Conversion Funnel Analysis).

**SELECT** 

COUNT(DISTINCT CASE WHEN page\_viewed IS NOT NULL THEN user\_id END) AS viewed,

COUNT(DISTINCT CASE WHEN added\_to\_cart = TRUE THEN user\_id END) AS added\_to\_cart,

COUNT(DISTINCT CASE WHEN purchased = TRUE THEN user\_id END) AS purchased FROM user\_activity;

10. Measure how much time users spend on the platform on average to evaluate engagement levels (Average Session Duration).

SELECT AVG(session\_duration\_seconds) AS avg\_session\_duration\_seconds FROM user\_activity;

11.Understand how many pages users view in a single session to measure content or product catalog effectiveness (Pages per Session).

SELECT session\_id, COUNT(page\_viewed) AS pages\_per\_session

FROM user\_activity
GROUP BY session\_id
ORDER BY pages\_per\_session DESC;

12. Analyze how frequently each user returns to the platform by counting sessions per user (Session Frequency per User).

SELECT user\_id, COUNT(DISTINCT session\_id) AS session\_count FROM user\_activity
GROUP BY user\_id
ORDER BY session\_count DESC;

13. Determine how many sessions result in just one page view to measure bounce rate and identify UX issues (Bounce Rate).

14.Track the percentage of users who click on any element after landing on a page to understand UI performance (Click-Through Rate – CTR).

SELECT COUNT(\*) FILTER (WHERE click\_event = TRUE) \* 100.0 / COUNT(\*) AS click\_through\_rate\_percent FROM user\_activity;

15. Calculate the proportion of users who added items to the cart but did not purchase to assess conversion leakage (Cart Abandonment Rate).

SELECT

COUNT(DISTINCT CASE WHEN added\_to\_cart = TRUE THEN user\_id END) AS added\_to\_cart,

COUNT(DISTINCT CASE WHEN purchased = TRUE THEN user\_id END) AS purchased,

(COUNT(DISTINCT CASE WHEN added\_to\_cart = TRUE THEN user\_id END) - COUNT(DISTINCT CASE WHEN purchased = TRUE THEN user\_id END)) \* 100.0 / NULLIF(COUNT(DISTINCT CASE WHEN added\_to\_cart = TRUE THEN user\_id END), 0) AS cart\_abandonment\_rate FROM user\_activity;

# 16. Measure the success rate of items added to the cart that led to a purchase to evaluate checkout performance (Cart-to-Purchase Rate).

#### **SELECT**

COUNT(\*) FILTER (WHERE added\_to\_cart = TRUE AND purchased = TRUE) \* 100.0 / NULLIF(COUNT(\*) FILTER (WHERE added\_to\_cart = TRUE), 0) AS cart\_to\_purchase\_rate\_percent FROM user\_activity;

# 17. Track how often users use coupons and how that translates into successful purchases to assess promotional impact (Coupon Conversion Rate).

#### SELECT

COUNT(\*) FILTER (WHERE coupon\_used = TRUE AND purchased = TRUE) \* 100.0 / NULLIF(COUNT(\*) FILTER (WHERE coupon\_used = TRUE), 0) AS coupon\_conversion\_rate FROM user\_activity;

### 18. Analyze how many users make repeat purchases to evaluate loyalty and customer retention (Repeat Purchase Rate – RPR).

```
SELECT COUNT(*) * 100.0 / NULLIF((SELECT COUNT(DISTINCT user_id) FROM user_activity WHERE purchased = TRUE), 0) AS repeat_purchase_rate_percent FROM (
    SELECT user_id
    FROM user_activity
    WHERE purchased = TRUE
    GROUP BY user_id
    HAVING COUNT(*) > 1
) AS repeat_users;
```

19. Calculate the total value of all completed purchases to measure gross merchandise value and overall sales performance (GMV – Gross Merchandise Value).

```
SELECT SUM(order_value) AS gross_merchandise_value
FROM user_activity
WHERE purchased = TRUE;
```

20. Determine the total revenue remaining after returns to understand true earnings and profitability (Net Revenue).

```
SELECT SUM(order_value) FILTER (WHERE purchased = TRUE AND returned = FALSE) AS net_revenue FROM user_activity;
```