



# **Operation Analytics and Investigating Metric Spike**

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***Case Study 2:  
Investigating Metric  
Spike***

# Investigating Metric Spike

## Description:

This project aims to investigate metric spikes in user engagement, growth, retention, and email interactions. The ultimate goal is to derive actionable insights that can inform strategic decisions to enhance user experience, optimize marketing efforts, and improve overall product engagement.

## SQL Task

### A. Weekly User Engagement:

Objective: Measure the activeness of users on a weekly basis.

Your Task: Write an SQL query to calculate the weekly user engagement.










#### Query:

```
SELECT
    DATE_FORMAT(occurred_at, '%Y-%u') AS week,
    user_id,
    COUNT(*) AS event_count
FROM
    events
GROUP BY
    DATE_FORMAT(occurred_at, '%Y-%u'), user_id
ORDER BY
    week;
```


#### Insights:

- Identifying users with consistent engagement can help target retention strategies and understand what drives active participation.
- Investigate the factors contributing to high engagement weeks and replicate successful strategies in other weeks.
- Develop personalized engagement campaigns for users with lower engagement levels.

SQL File 3\*SQL File 4\*SQL File 5\*SQL File 5\*SQL File 6\* x





Limit to 1000 rows




```
1  -- A. Weekly User Engagement
2
3  •  SELECT
4      DATE_FORMAT(occurred_at, '%Y-%u') AS week,
5      user_id,
6      COUNT(*) AS event_count
7  FROM
8      events
9  GROUP BY
10     DATE_FORMAT(occurred_at, '%Y-%u'), user_id
11  ORDER BY
12     week;
```


<

Result Grid



Filter Rows:

Export: 

Wrap Cell Content: 

	week	user_id	event_count
▶	2014-18	8	2
	2014-18	22	8
	2014-18	66	14
	2014-18	163	22
	2014-18	172	15
	2014-18	209	9
	2014-18	232	47
	2014-18	239	9
	2014-18	253	24
	2014-18	287	16
	2014-18	319	10
	2014-18	365	7
	2014-18	373	8
	2014-18	404	3
	2014-18	434	5
	2014-18	441	28
	2014-18	492	19
	2014-18	576	7
	2014-18	589	11




Result 2 x

## B. User Growth Analysis:

Objective: Analyze the growth of users over time for a product.

Your Task: Write an SQL query to calculate the user growth for the product.

```
1  -- B. User Growth Analysis
2  •  SELECT
3      DATE_FORMAT(created_at, '%Y-%m') AS month,
4      COUNT(*) AS new_users
5  FROM
6      users
7  GROUP BY
8      DATE_FORMAT(created_at, '%Y-%m')
9  ORDER BY
10     month;
```

Result Grid |  Filter Rows:  | Export:  | Wrap Cell Content: 

	month	new_users
▶	2013-01	160
	2013-02	160
	2013-03	150
	2013-04	181
	2013-05	214
	2013-06	213
	2013-07	284
	2013-08	316
	2013-09	330
	2013-10	390
	2013-11	399
	2013-12	486
	2014-01	552
	2014-02	525
	2014-03	615
	2014-04	726
	2014-05	779
	2014-06	873
	2014-07	997
	2014-08	1031

Query:

```
SELECT
    DATE_FORMAT(created_at, '%Y-%m') AS month,
    COUNT(*) AS new_users
FROM
    users
GROUP BY
    DATE_FORMAT(created_at, '%Y-%m')
ORDER BY
    month;
```

Insights:

- Continuous monitoring of user growth trends helps understand the impact of marketing campaigns or product changes on user acquisition.
- Analyze marketing efforts or product features introduced in January and February that led to higher growth.
- Investigate reasons for the slight decline in March and address any issues to sustain growth.

### C. Weekly Retention Analysis:

Objective: Analyze the retention of users on a weekly basis after signing up for a product.

Your Task: Write an SQL query to calculate the weekly retention of users based on their sign-up cohort.

Query:

```
WITH cohort AS (
    SELECT
        user_id,
        DATE_FORMAT(created_at, '%Y-%u') AS signup_week
    FROM
        users
), weekly_engagement AS (
```

```

SELECT
    user_id,

    DATE_FORMAT(occurred_at, '%Y-%u') AS active_week
FROM
    events
GROUP BY
    user_id, active_week
)
SELECT
    c.signup_week,
    we.active_week,
    COUNT(DISTINCT we.user_id) AS retained_users
FROM
    cohort c
JOIN
    weekly_engagement we ON c.user_id = we.user_id
GROUP BY
    c.signup_week, we.active_week
ORDER BY
    c.signup_week, we.active_week;

```

## Insights:

- Trends indicate that while some users remain active long-term, retention generally declines over time, highlighting the need for ongoing engagement efforts.
- Investigate successful onboarding and engagement strategies for high-retention cohorts.
- Develop retention programs targeting users at risk of churn to improve long-term retention rates.
- By tracking the number of users who remain active each week following their sign-up to understand retention rates and identify successful onboarding strategies.

SQL File 3\*   SQL File 4\*   SQL File 5\*   SQL File 5\*   SQL File 6\* x

Limit to 1000 rows

```

13  -- C. Weekly Retention Analysis
14
15  WITH cohort AS (
16      SELECT
17          user_id,
18          DATE_FORMAT(created_at, '%Y-%u') AS signup_week
19      FROM
20          users
21  ), weekly_engagement AS (
22      SELECT
23          user_id,
24          DATE_FORMAT(occurred_at, '%Y-%u') AS active_week
25      FROM
26          events
27      GROUP BY
28          user_id, active_week
29  )
30  SELECT
31      c.signup_week,
32      we.active_week,
33      COUNT(DISTINCT we.user_id) AS retained_users
34  FROM cohort c
35  JOIN weekly_engagement we ON c.user_id = we.user_id
36  GROUP BY c.signup_week, we.active_week
37  ORDER BY c.signup_week, we.active_week;
38

```

Result Grid | Filter Rows: | Export: | Wrap Cell Content: | Fetch r

	signup_week	active_week	retained_users
▶	2013-01	2014-18	2
	2013-01	2014-19	3

## D. Weekly Engagement Per Device:

Objective: Measure the activeness of users on a weekly basis per device.

Your Task: Write an SQL query to calculate the weekly engagement per device.

45	-- D. Weekly Engagement Per Device
46	
47	• SELECT
48	DATE_FORMAT(occurred_at, '%Y-%u') AS week,
49	device,
50	COUNT(*) AS event_count
51	FROM
52	events
53	GROUP BY
54	DATE_FORMAT(occurred_at, '%Y-%u'), device
55	ORDER BY
56	week;
57	

Result Grid	Filter Rows:	Export:	Wrap Cell
week	device	event_count	
2014-18	acer aspire desktop	71	
2014-18	acer aspire notebook	215	
2014-18	amazon fire phone	84	
2014-18	asus chromebook	286	
2014-18	dell inspiron desktop	198	
2014-18	dell inspiron notebook	569	
2014-18	hp pavilion desktop	141	
2014-18	htc one	192	
2014-18	ipad air	358	
2014-18	ipad mini	229	
2014-18	iphone 4s	219	
2014-18	iphone 5	774	
2014-18	iphone 5s	512	
2014-18	kindle fire	64	
2014-18	lenovo thinkpad	858	
2014-18	mac mini	64	
2014-18	macbook air	538	
2014-18	macbook pro	1705	

Query:

```

SELECT
    DATE_FORMAT(occurred_at, '%Y-%u') AS week,
    device,
    COUNT(*) AS event_count
FROM
    events
GROUP BY

```



```
DATE_FORMAT(occurred_at, '%Y-%u'), device
ORDER BY
week;
```

Insights:

- Lower engagement on devices like the Amazon Fire Phone suggests potential usability issues or lack of optimization.
- Focus on optimizing the user experience for popular devices to maintain high engagement levels.
- Investigate and address usability issues on less popular devices to improve overall engagement.

## E. Email Engagement Analysis:

objective: Analyze how users are engaging with the email service.

Your Task: Write an SQL query to calculate the email engagement metrics.

Query:

```
SELECT
    DATE_FORMAT(occurred_at, '%Y-%u') AS week,
    action,
    COUNT(*) AS action_count
FROM
    mailevents
GROUP BY
    DATE_FORMAT(occurred_at, '%Y-%u'), action
ORDER BY
    week;
```

Insights:

- High email open rates suggest effective subject lines and timing, while click-through rates indicate the relevance of email content.
- Continue refining email strategies based on open and click-through rates to enhance engagement.
- Experiment with different email content, subject lines, and send times to optimize email performance.
- Calculate email engagement metrics, such as open and click-through rates, to assess the effectiveness of email campaigns and user interactions with email content.

```

58  -- E. Email Engagement Analysis
59  •  SELECT
60      DATE_FORMAT(occurred_at, '%Y-%u') AS week,
61      action,
62      COUNT(*) AS action_count
63  FROM
64      emailevents
65  GROUP BY
66      DATE_FORMAT(occurred_at, '%Y-%u'), action
67  ORDER BY
68      week;

```

&lt;

Result Grid



Filter Rows:

Export:



Wrap Cell Content:



	week	action	action_count
►	2014-18	email_clickthrough	187
	2014-18	email_open	332
	2014-18	sent_reengagement_email	98
	2014-18	sent_weekly_digest	908
	2014-19	email_clickthrough	434
	2014-19	email_open	919
	2014-19	sent_reengagement_email	164
	2014-19	sent_weekly_digest	2602
	2014-20	email_clickthrough	479
	2014-20	email_open	971
	2014-20	sent_reengagement_email	175
	2014-20	sent_weekly_digest	2665
	2014-21	email_clickthrough	498
	2014-21	email_open	995
	2014-21	sent_reengagement_email	179
	2014-21	sent_weekly_digest	2733
	2014-22	email_clickthrough	453
	2014-22	email_open	1026
	2014-22	sent_reengagement_email	179
	2014-22	sent_weekly_digest	2822

## Conclusion:

The analysis provides valuable insights into user behavior and engagement, highlighting periods of high activity, growth trends, retention patterns, device preferences, and email interactions. These insights can inform strategic decisions to enhance user experience, optimize marketing efforts, and improve overall product engagement. By continuously monitoring and analyzing these metrics, the product team can make data-driven decisions to sustain and enhance user engagement and satisfaction.