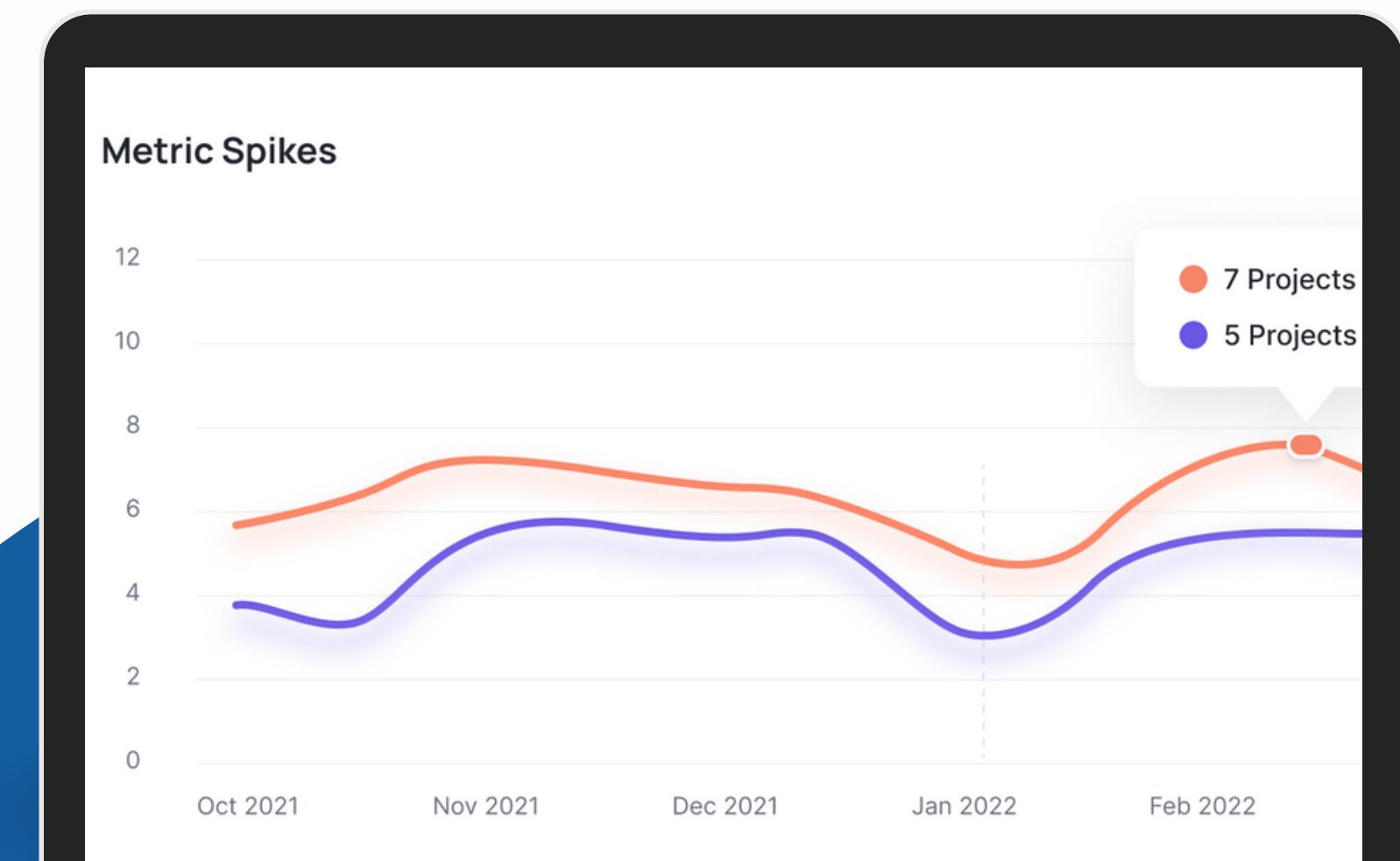
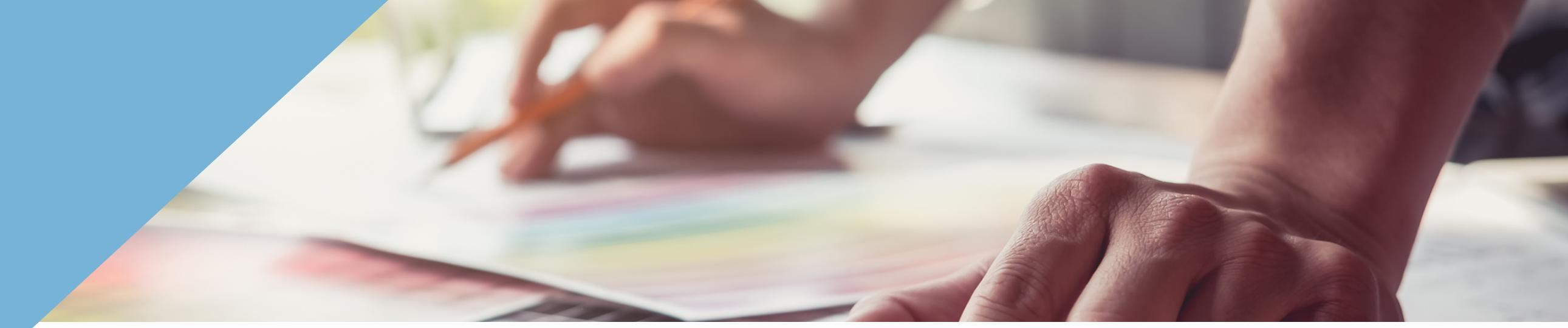


# OPERATION ANALYTICS AND INVESTIGATING METRIC SPIKE

By : Iswariya S



# Agenda

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# PROJECT DESCRIPTION

The Operation Analytics and Investigating Metric Spike Data Analytics project is a strategic initiative designed to empower organizations with enhanced insights into their operational data. Focused on real-time monitoring and analysis, the project employs advanced data analytics techniques to detect and investigate metric spikes efficiently. By leveraging cutting-edge technologies and algorithms, the system aims to provide actionable intelligence for prompt decision-making. The project encompasses the development of a comprehensive analytics platform that integrates seamlessly with existing systems, ensuring a streamlined approach to data collection, processing, and interpretation. With a primary goal of optimizing operational performance, this project is poised to deliver valuable outcomes for businesses seeking to proactively address issues, improve efficiency, and make informed strategic decisions based on a thorough understanding of their metrics and data patterns.



# APPROACH

The project employs advanced data analytics to monitor real-time operational metrics, detect irregularities, and derive insights for informed decision-making.



- **Established the database structure and tables**

Developed a database and subsequently constructed tables based on the provided structure and interconnections.

- **Conducted an in-depth analysis**

Employed SQL to thoroughly analyze the dataset, addressing the posed questions. After reviewing the provided dataset, tables were created to compute diverse queries.

- **Integrated Data for Informed Decision-Making**

Combined data fragments, organized the table structure for extracting business insights, retrieved the necessary outcomes, and thereby generated valuable insights for the company, enabling informed and strategic decision-making.

## TECH-STACK USED

The project encompasses MySQL Community Server for database management, MS Excel for data analysis, and MS PowerPoint for effective insights presentation, providing a well-rounded toolkit for streamlined data analytics and decision-making.



1

Used MySQL Community server - (GPL Version 8.0.34 Connector Version C++ 8.0.34 For creating database)

2

Microsoft Excel - ( For import and Export of data)

3

Microsoft PowerPoint - ( For Presentation )

# INSIGHTS AND KNOWLEDGE

## Case Study 1: Job Data Analysis

### 1.Jobs Reviewed Over Time:

- Insights: Identify peak review periods.
- Knowledge: Forecast job review trends.

### 2.Throughput Analysis:

- Insights: Assess efficiency and detect bottlenecks.
- Knowledge: Improve process efficiency.

### 3.Language Share Analysis:

- Insights: Understand language distribution.
- Knowledge: Tailor services based on language popularity.

### 4.Duplicate Rows Detection:

- Insights: Ensure data integrity.
- Knowledge: Handle duplicate records for cleaner analysis.

## Case Study 2: Investigating Metric Spike

### 1.Weekly User Engagement:

- Insights: Recognize user behavior patterns.
- Knowledge: Optimize engagement during peak periods.

### 2.User Growth Analysis:

- Insights: Track and understand user acquisition.
- Knowledge: Sustain and accelerate user growth.

### 3.Weekly Retention Analysis:

- Insights: Evaluate user retention factors.
- Knowledge: Improve retention through engagement.

### 4.Weekly Engagement Per Device:

- Insights: Analyze user behavior across devices.
- Knowledge: Optimize for specific platforms.

### 5.Email Engagement Analysis:

- Insights: Assess email communication effectiveness.
- Knowledge: Improve email engagement for better user interaction.



# RESULT

## Case Study 1: Job Data Analysis

### A. Jobs Reviewed Over Time

SQL query to calculate the number of jobs reviewed per hour for each day in November 2020.

```
#TASK-1(Jobs Reviewed Over Time)
• select * from job_data;
• select avg(t) as 'avg jobs reviewed per day per hour',
  avg(p) as 'avg jobs reviewed per day per second'
from
(select
ds,
((count(job_id)*3600)/sum(time_spent)) as t,
((count(job_id))/sum(time_spent)) as p
from
job_data
where
month(ds)=11
group by ds) a;
```



**avg jobs reviewed per  
day per hour**

**126.18048333**

**avg jobs reviewed per  
day per second**

**0.03505000**



## B. Throughput Analysis

SQL query to calculate the 7-day rolling average of throughput (number of events per second).

- If the goal is to understand short-term variations and daily patterns, using the "Daily Throughput" metric is appropriate.
- If the goal is to identify broader trends and patterns while reducing the impact of daily noise, the "7-Day Rolling Average" is a better choice.

It's often beneficial to use both metrics in conjunction to gain a comprehensive understanding of throughput trends.

```
SELECT ROUND(COUNT(event)/SUM(time_spent), 2) AS "Weekly Throughput" FROM job_data;  
  
SELECT ds AS Dates, ROUND(COUNT(event)/SUM(time_spent), 2) AS "Daily Throughput" FROM job_data  
GROUP BY ds ORDER BY ds;
```

### For Daily Throughput:

Dates	Daily Throughput
2020-11-25	0.02
2020-11-26	0.02
2020-11-27	0.01
2020-11-28	0.06
2020-11-29	0.05
2020-11-30	0.05

### For Weekly Throughput:

Weekly Throughput
0.03

## C. Language Share Analysis

SQL query to calculate the percentage share of each language over the last 30 days.

```
SELECT language AS Languages, ROUND(100 * COUNT(*)/TOTAL, 2) AS Percentage, sub.total  
FROM job_data  
CROSS JOIN (SELECT COUNT(*) AS total FROM job_data) AS sub  
GROUP BY language, sub.total;
```

	Languages	Percentage	total
▶	English	12.50	16
	Arabic	12.50	16
	Persian	37.50	16
	Hindi	12.50	16
	French	12.50	16
	Italian	12.50	16

## D. Duplicate Rows Detection

SQL query to display duplicate rows from the job\_data table.

```
SELECT actor_id, COUNT(*) AS Duplicates FROM job_data  
GROUP BY actor_id HAVING COUNT(*) > 1;
```

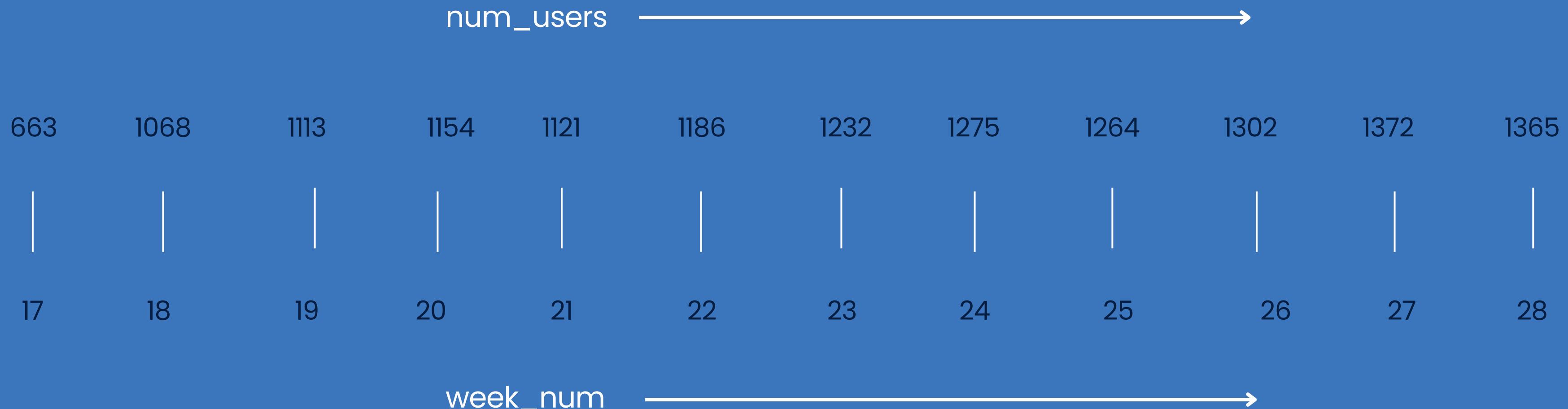
	actor_id	Duplicates
▶	1001	2
	1006	2
	1003	4
	1005	2
	1002	2
	1007	2
	1004	2

# Case Study 2: Investigating Metric Spike

## A. Weekly User Engagement

SQL query to calculate the weekly user engagement.

```
select extract(week from occurred_at) as week_number,  
       count(distinct user_id) as active_user  
  from events_tb1  
 where event_type='engagement'  
 group by week_number  
 order by week_number
```



## B. User Growth Analysis

SQL query to calculate the user growth for the product.

```
select year, week_num, num_users, sum(num_users)
over (order by year, week_num) as cum_users
from (
  select extract(year from created_at) as year, extract(week from created_at) as week_num, count(distinct user_id) as num_users
  from users_tb1
  where state='active'
  group by year, week_num
  order by year, week_num)sub
```

	year	week_num	num_users	cum_users
▶	2013	0	23	23
	2013	1	30	53
	2013	2	48	101
	2013	3	36	137
	2013	4	30	167
	2013	5	48	215
	2013	6	38	253
	2013	7	42	295
	2013	8	34	329
	2013	9	43	372
	2013	10	32	404
	2013	11	31	435
	2013	12	33	468

## C. Weekly Retention Analysis

SQL query to calculate the weekly retention of users based on their sign-up cohort.

```
with cte1 as (
  select distinct user_id,
    Extract (week from occurred at) as signup_week
  from events_tb1
  where event_type = 'signup_flow'
  and event_name = 'complete_signup' and extract (week from occurred at) = 18 ),
  cte2 as (select distinct user_id,
    Extract (week from occurred at) as engagement_week
  from events_tb1
  where event_type = 'engagement')
  select count(user_id) total_engaged_users,
    sum(case when retention_week > 8 then 1 else end) as retained_users
  from (select a.user_id, a.signup_week,
    b.engagement_week, b.engagement_week-a.signup_week as retention_week
  from cte1 a
  LEFT JOIN cte2 b
  on a.user_id = b.user_id
  order by a.user_id) sub
```

total\_engaged\_users

317

retained\_users

236

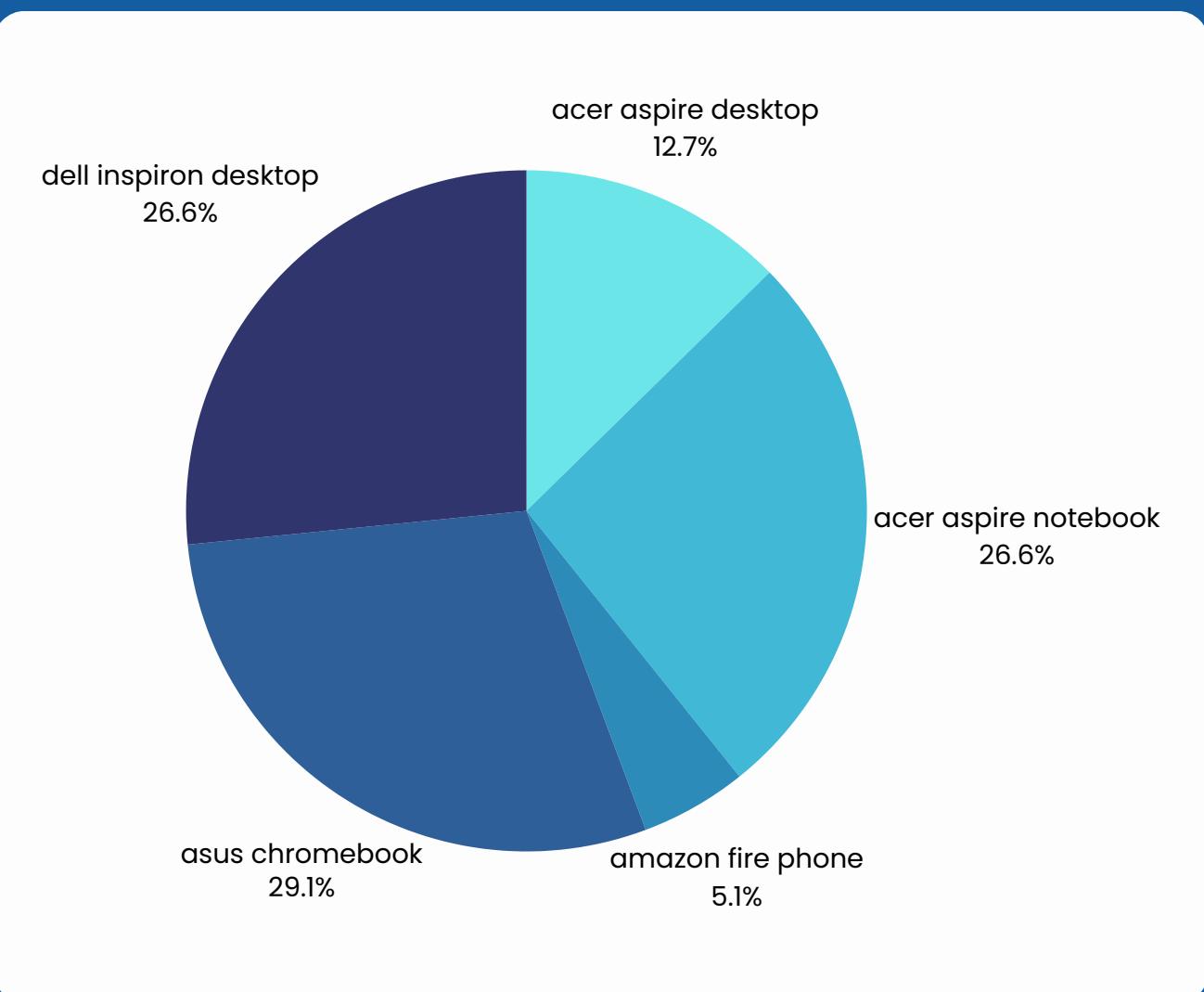


## D. Weekly Engagement Per Device

SQL query to calculate the weekly engagement per device.

```
with cte as (select extract (year from occurred_at)||'-'||extract(week from occurred_at) as weeknum  
device, count(distinct user_id) as usercnt  
from events_tb1  
where event_type = 'engagement'  
group by weeknum, device  
order by weeknum)  
select weeknum, device, usercnt  
from cte
```

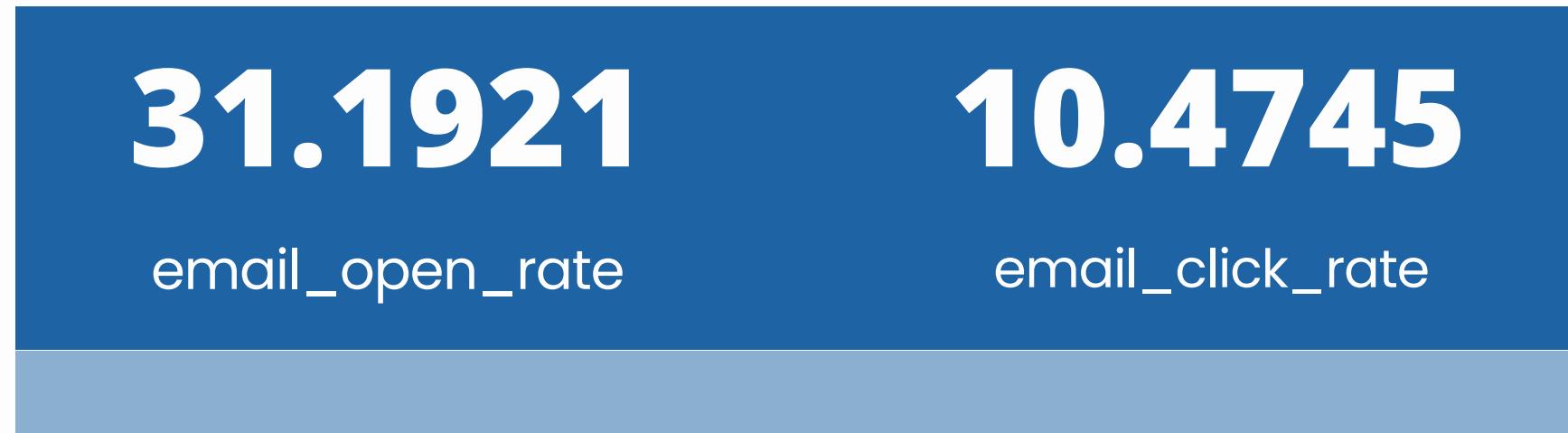
weeknum	device	usercnt
2014-18	acer aspire desktop	10
2014-18	acer aspire notebook	21
2014-18	amazon fire phone	4
2014-18	asus chromebook	23
2014-18	dell inspiron desktop	21



## E. Email Engagement Analysis

SQL query to calculate the email engagement metrics.

```
select
  100 * sum(case when email_cat = 'email_open' then 1 else 0 end) /
  sum(case when email_cat = 'email sent' then 1 else 0 end) as email_open_rate,
  100 * sum(case when email_cat = 'email clicked' then 1 else 0 end) /
  sum(case when email_cat = 'email sent' then 1 else 0 end) as email_click_rate
from (select*,
  case
    when action in ('sent_weekly_digest', 'sent_reengagement_email') then 'email_sent'
    when action in ('email_open') then 'email_open'
    when action in ('email_clickthrough') then 'email_clicked'
  end as email_cat
  from email_events) sub
```



# CONCLUSION

In this project, I learned advanced SQL techniques like Windows Functions, improved my SQL skills, and discovered how to ask the right questions for business growth. I gained insights into finding valuable information, understanding areas for improvement in companies, and investigating changes in metrics.