OPERATION ANALYTICS AND INVESTIGATING METRIC SPIKE

Advanced SQL

- ♣ Project Description: This project involves analysing a company's end-to-end operations. Operational Analytics is a crucial process that helps identify areas for improvement within the company. As a Data Analyst, I have worked closely with various teams, such as operations, support, and marketing, helping them derive valuable insights from the data they collect. One of the key aspects of Operational Analytics is investigating metric spikes. This involves understanding and explaining sudden changes in key metrics, such as a dip in daily user engagement or a drop in sales. As a Data Analyst, you'll need to answer these questions daily, making it crucial to understand how to investigate these metric spikes.
- ♣ Approach: As a lead Data Analyst these following steps were taken to analyse the data and provide valuable insights that would help improve the company's operations and understand sudden changes in key metrics.
 - Create Database and Tables: First off, all a new database named Project3 was
 created and all the tables (Both for case study 1 and case study 2) provided in MyExcel csv format were imported in MYSQL workbench by following below query.

```
LOAD DATA INFILE 'C:/ProgramData/MySQL/MySQL Server 8.0/Uploads/users.csv'
INTO TABLE users
FIELDS TERMINATED BY ','
ENCLOSED BY '"'
LINES TERMINATED BY '\n'
IGNORE 1 ROWS; -- Skip the header row if present
```

- 2. **Formulate queries**: Determined the specific questions I wanted to answer and tried to understand what result this query wants. Formulated SQL queries that retrieve the relevant data to answer those questions. Started with simple queries and gradually build more complex ones as needed.
- 3. **Write Queries**: Used the SQL editor in MySQL Workbench to write and execute SQL queries. Ensured that the queries are accurate, efficient, and properly structured to extract the required information from the database.
- 4. **Execute Queries**: Ran SQL queries against the database. Reviewed the query results to ensure they match with question expectations. Again, Refined queries as necessary to get the desired results.

- **Tech-Stack Used**: In this project, MySQL Workbench 8.0 was used as a tool for creating databases, tables and writing SQL queries. To perform Operation Analytics and Investigating Metric Spike and answer questions posed by the management team.
 - Microsoft Excel was used for importing data.
 - Microsoft Word for Preparing the Project Report.
- **↓ INSIGHT**: While working on this project I gained hands on experience with real life use cases. I gained knowledge on Advanced SQL queries and used it to answer some really interesting questions. For this project two case studies were given.

Case Study 1: Job Data Analysis

Table named job_data with the following columns:

- job_id: Unique identifier of jobs
- actor_id: Unique identifier of actor
- event: The type of event (decision/skip/transfer).
- language: The Language of the content
- time_spent: Time spent to review the job in seconds.
- org: The Organization of the actor
- ds: The date in the format yyyy/mm/dd (stored as text).



Table 1: Job_data

Tasks:

A. Jobs Reviewed Over Time:

- Objective: Calculate the number of jobs reviewed per hour for each day in November 2020.
- SQL Task: Write an SQL query to calculate the number of jobs reviewed per hour for each day in November 2020.

```
SELECT

ds AS Job_Date,

COUNT(job_id) AS Jobs_Per_Day,

SUM(time_spent) AS Total_Time_Spent_in_seconds,

ROUND((COUNT(job_id)*3600) / SUM(Time_Spent)/3600) AS Jobs_Reviewed_per_Hour_Day

FROM

job_data

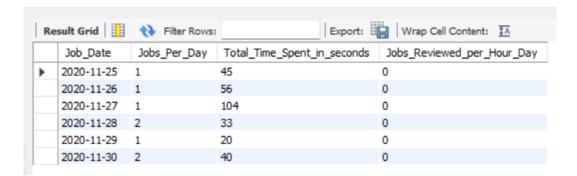
WHERE

ds BETWEEN '2020-11-01' AND '2020-11-30'

GROUP BY Job_Date

ORDER BY Job_Date;
```

SQL QUERY



OUTPUT

INSIGHT: For each day in November 2020 no. of. Jobs reviewed per hour is 0.

B. Throughput Analysis:

- Objective: Calculate the 7-day rolling average of throughput (number of events per second).
- SQL Task: Write an SQL query to calculate the 7-day rolling average of throughput.
 Additionally, explain whether you prefer using the daily metric or the 7-day rolling average for throughput, and why.

```
SELECT

ds, throughput,

AVG(throughput) OVER (ORDER BY ds ROWS BETWEEN 6 PRECEDING AND CURRENT ROW)

AS

rolling_avg_events_per_second

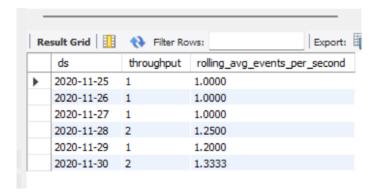
FROM (

SELECT ds, COUNT(*) AS throughput

FROM job_data

GROUP BY ds
) AS daily_events;
```

SQL QUERY



OUTPUT

INSIGHT: The rolling average throughput is calculated by taking the average of throughput values over a sliding window of a specified number of days. In this case, we're using a 7-day rolling average, which means for each row in the result, the average is calculated over the current day and the preceding 6 days.

So, for each row in the result set, the rolling average throughput is calculated by summing the throughput values for the current day and the preceding 6 days, and then dividing that sum by 7.

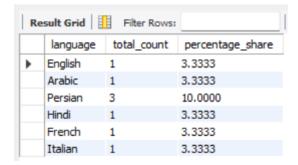
This rolling average provides a smoothed average over a specified period, helping to identify trends or patterns in the data while reducing the effects of short-term fluctuations.

C. Language Share Analysis:

- Objective: Calculate the percentage share of each language in the last 30 days.
- SQL Task: Write an SQL query to calculate the percentage share of each language over the last 30 days.

```
SET @num_days := 30;
    WITH no_of_days AS (
    SELECT
        language,
        COUNT(*) AS total_count
    FROM
        job_data
    WHERE
        ds <= CURDATE() - INTERVAL @num_days DAY
    GROUP BY
        language
)
SELECT
    language,
    total_count,
    (total_count * 100) / @num_days AS percentage_share
FROM
    no_of_days ;
```

SQL QUERY



OUTPUT

INSIGHT: To calculate the percentage share of each language over the last 30 days in SQL, aggregate the data by language and then calculate the percentage share based on the count of each language occurrence within the last 30 days compared to the total count of all languages within the same period.

Persian language has the highest percentage share among all other languages.

D. Duplicate Rows Detection:

- Objective: Identify duplicate rows in the data.
- SQL Task: Write an SQL query to display duplicate rows from the job_data table.

```
SELECT *

FROM job_data

WHERE job_id IN (

SELECT job_id

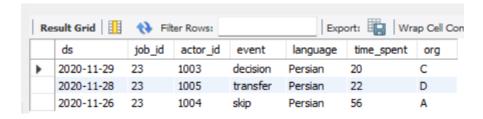
FROM job_data

GROUP BY job_id

HAVING COUNT(*) > 1

);

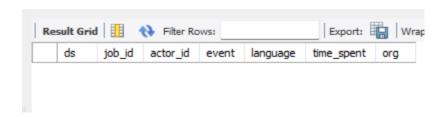
SQL Query
```



OUTPUT: Duplicate rows using job_id column

```
SELECT *
FROM job_data
GROUP BY ds, job_id, actor_id, event, language, time_spent, org
HAVING COUNT(*)>1;
```

SQL Query: Duplicate using entire column



OUTPUT: No rows are duplicate in entire row

INSIGHT: There were no rows duplicate in the entire columns. But based on job_id columns 3 rows were found to be duplicate.

Case Study 2: Investigating Metric Spike

Following are the three tables:

users: Contains one row per user, with descriptive information about that user's account.

events: Contains one row per event, where an event is an action that a user has taken (e.g., login, messaging, search).

email events: Contains events specific to the sending of emails.

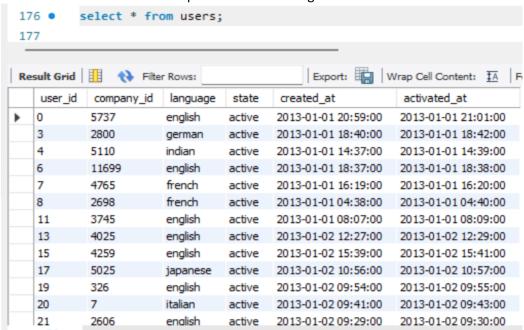


Table 1: Portion of Users table

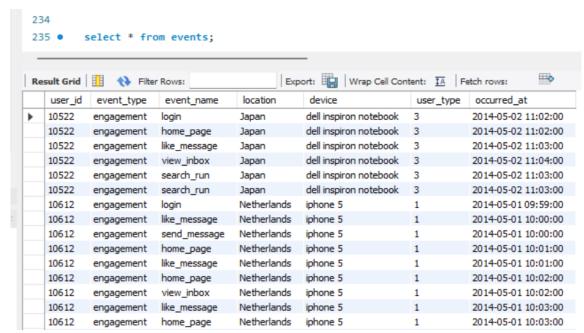


Table 2: Portion of Events table

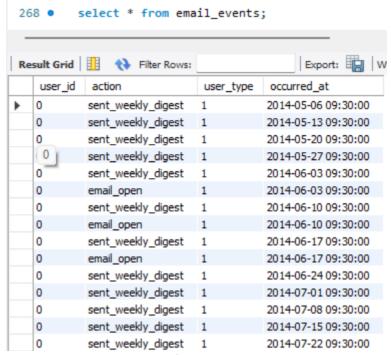


Table 3: Portion of Email_events table.

Tasks:

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.

```
week_number
                                                                             active_user_count
SELECT
    extract(week from occurred_at)
                                                                             1068
                                                                18
    AS week_number,
                                                                 19
                                                                             1113
    COUNT( DISTINCT user_id) AS active_user_count
                                                                20
                                                                             1154
                                                                             1121
FROM
                                                                22
                                                                             1186
    events
                                                                             1232
                                                                23
    where event_type='engagement'
                                                                24
                                                                             1275
GROUP BY
                                                                25
                                                                             1264
                                                                26
                                                                             1302
    week_number
                                                                27
                                                                             1372
ORDER BY
                                                                28
                                                                             1365
                                                                                               32
                                                                                                          1225
    week_number;
                                                                29
                                                                             1376
                                                                                                          1225
                                                                30
                                                                             1467
                                                                                               34
                                                                                                          1204
                                                                31
                                                                             1299
```

SQL Query OUTPUT

INSIGHT: Measuring the activeness of users on a weekly basis typically involves determining how many unique users were active during each week. "Active" users are those who performed some form of activity within the given week.

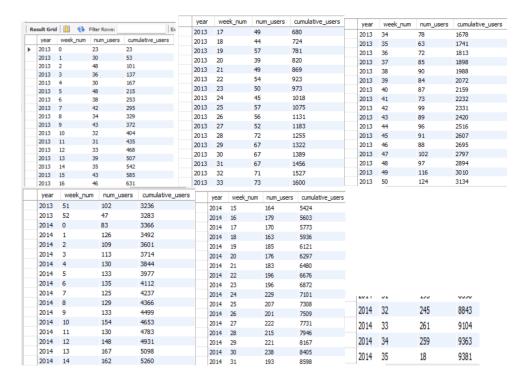
Based on the given data users were highly active in week 30 and least active in week 35.

B. User Growth Analysis:

- Objective: Analyse the growth of users over time for a product.
- Your Task: Write an 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 cumulative_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
group by year,week_num
order by year,week_num) sub;
```

SQL Query and OUTPUT



INSIGHT: To calculate the user growth for a product, the total number of users between two different periods of time were compared.

The above query gives the total number of users registered in week of the year, as well as the cumulative number of users up to that week. We can then analyse the growth trend over time based on this data. From the output we observe that for some week users increases and in some week user decreases.

C. Weekly Retention Analysis:

- Objective: Analyse 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.

```
WITH ctel AS (
       SELECT DISTINCT
           user id AS signed users,
           EXTRACT(WEEK FROM occurred_at) AS signup_week
       WHERE
           event_type = 'signup_flow' AND event_name = 'complete_signup'
   ),
   cte2 A5 (
       SELECT distinct
           user_id,
           EXTRACT(WEEK FROM occurred_at) AS engagement_week
       FROM
       WHERE
           event_type = 'engagement'
 SELECT
     signup_week,
     COUNT(signed_users) AS total_signed_users,
     SUM(CASE WHEN retention_week > 0 THEN 1 ELSE 0 END) AS week1_retained_users,
     SUM(CASE WHEN retention_week > 0 THEN 1 ELSE 0 END) / COUNT(signed_users) *100 as week1_retention_rate
FROM (
     SELECT
        a.signed users.
        a.signup week,
        b.engagement_week,
        b.engagement week - a.signup week AS retention week
     LEFT JOIN
       cte2 b ON a.signed_users = b.user_id
 GROUP BY
     signup_week;
```

SQL QUERY

_	signup_week	total cionad usars	week1_retained_users	week1_retention_rate		1	1	
_					,		1	
۰	17	278	206	74.1007				
	18	615	452	73.4959				
	19	677	492	72.6736				
	20	682	506	74.1935				
	21	644	461	71.5839				
	22	694	498	71.7579				
	23	707	511	72.2772				
	24	700	471	67.2857				
	25	671	464	69.1505				
	26	636	435	68.3962				
	27	697	475	68.1492				
	28	596	381	63.9262				
	29	588	367	62.4150				
	30	614	376	61,2378				
	31	451	258	57.2062				
	32	508	263	51.7717		34		
	33	456	195	42.7632		35	35 18	35 18 0

INSIGHT: Weekly retention refers to the percentage of users who continue to engage with a product or service over time, specifically on a weekly basis, after initially signing up or starting to use the product. It measures the ability of a product to retain its users over consecutive weeks. This query weekly retention rate after first week for each sign-up cohort.

This analysis allows businesses to track the effectiveness of their retention strategies and assess user satisfaction and loyalty.

We observe that in week 34 approximately 14% are retained where as in week 20 approximately 74% of total users are retained.

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.

```
SELECT

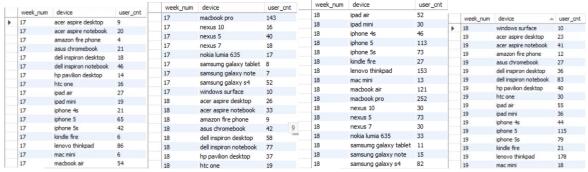
EXTRACT(WEEK FROM occurred_at) AS week_num,
device,
COUNT(DISTINCT user_id) AS user_cnt

FROM
events

WHERE
event_type = 'engagement'

GROUP BY
EXTRACT(WEEK FROM occurred_at), device

ORDER BY
week_num, device;
```



^{*}more rows are there which are not shown here

OUTPUT

INSIGHTS: This query gives the weekly engagement count for each device. In week 30 device MacBook pro was used by highest no. of users that is 322.

E. Email Engagement Analysis:

- Objective: Analyse how users are engaging with the email service.
- Your Task: Write an SQL query to calculate the email engagement metrics.

```
SELECT

DATE(occurred_at) AS date,

COUNT(*) AS total_emails_sent,

COUNT(DISTINCT user_id) AS unique_recipients,

SUM(CASE WHEN action = 'email_open' THEN 1 ELSE 0 END) AS emails_opened,

SUM(CASE WHEN action = 'email_clickthrough' THEN 1 ELSE 0 END) AS emails_clicked,

SUM(CASE WHEN action IN ('sent_weekly_digest', 'sent_reengagement_email') THEN 1 ELSE 0 END)

AS emails_sent

FROM

email_events

GROUP BY

DATE(occurred_at);
```

SQL QUERY

	date	total_emails_sent	unique_recipients	emails_opened	emails_clicked	emails_ser
-	2014-05-01	680	474	145	61	474
	2014-05-02	704	480	142	82	480
	2014-05-03	73	27	23	23	27
	2014-05-04	68	25	22	21	25
	2014-05-05	1164	794	255	115	794
	2014-05-06	757	507	168	82	507
	2014-05-07	647	443	141	63	443
	2014-05-08	709	488	156	65	488
	2014-05-09	687	475	148	64	475
	2014-05-10	69	27	22	20	27
	2014-05-11	86	32	29	25	32
	2014-05-12	1174	799	258	117	799
	2014-05-13	807	522	193	92	522
	2014-05-13	667	454	151	62	454
	2014-05-14	703	500	146	57	500
	2014-05-15	750	496	162	92	496
	2014-05-16	100		33	32	35
_	1		35			
_	date	total_emails_sent	unique_recipients	emails_opened	emails_clicked	emails_sent
	2014-05-18	89	34	28	27	34
	2014-05-19	1244	816	282	146	816
	2014-05-20	845	544	204	97	544
	2014-05-21	668	466	142	60	466
	2014-05-22	719	510	146	63	510
	2014-05-23	766	516	168	82	516
	2014-05-24	104	38	34	32	38
	2014-05-25	59	22	19	18	22
	2014-05-26	1255	837	285	133	837
	2014-05-27	829	556	187	86	556
	2014-05-28	673	480	145	48	480
	2014-05-29	764	532	176	56	532
	2014-05-30	776	527	173	76	527
	2014-05-31	87	32	29	26	32
	2014-06-01	96	37	31	28	37
	2014-06-02	1228	855	256	117	855
_	2014-06-03	873	577	195	101	577
	2014-06-04	704	494	148	62	494
	2014-06-05	788	554	158	76	554
	2014-06-06	812	557	173	82	557
	2014-06-07	77	29	26	22	29
	2014-06-08	113	44	37	32	44
	2014-06-09	1272	878	264	130	878
	2014-06-10	889	593	201	95	593
	2014-06-11	720	509	149	62	509
	2014-06-12	858	573	198	87	573
	2014-06-13	883	575	200	108	575
	2014-06-14	78	28	26	24	28
	2014-06-15	96	37	32	27	37
L	2014-06-16	1364	917	305	142	917
	2014-06-17	915	605	209	101	605
L	2014-06-18	759	528	169	62	528
	2014-06-19	877	598	189	90	598
	2014-06-20	894	597	205	92	597

INSIGHT: the above query calculates various email engagement metrics such as total emails sent, unique recipients, emails opened, emails clicked, and emails sent for each date.