

Project 3- Operation Analytics and Investigating Metric Spike

Project description:

Operation Analytics is the analysis done for the complete end to end operations of a company. With the help of this, the company then finds the areas on which it must improve upon. You work closely with the ops team, support team, marketing team, etc and help them derive insights out of the data they collect. Investigating metric spike is also an important part of operation analytics as being a Data Analyst you must be able to understand or make other teams understand questions like- Why is there a dip in daily engagement? Why have sales taken a dip? Etc. Questions like these must be answered daily and for that its very important to investigate metric spike.

In this project I am working as data analyst lead and am required to answer certain questions pertaining to operation analytics. I will be preparing datasets and then study and run queries on it to come up with answers.

Approach:

My approach towards the completion of this project is simple. At first I will complete the dataset on excel and save it as a csv file and then use MySQL workbench and import this file as dataset where I will attend every question and run queries on the data and come up with answers and insights.

Tech stack used:

For the project, I used:

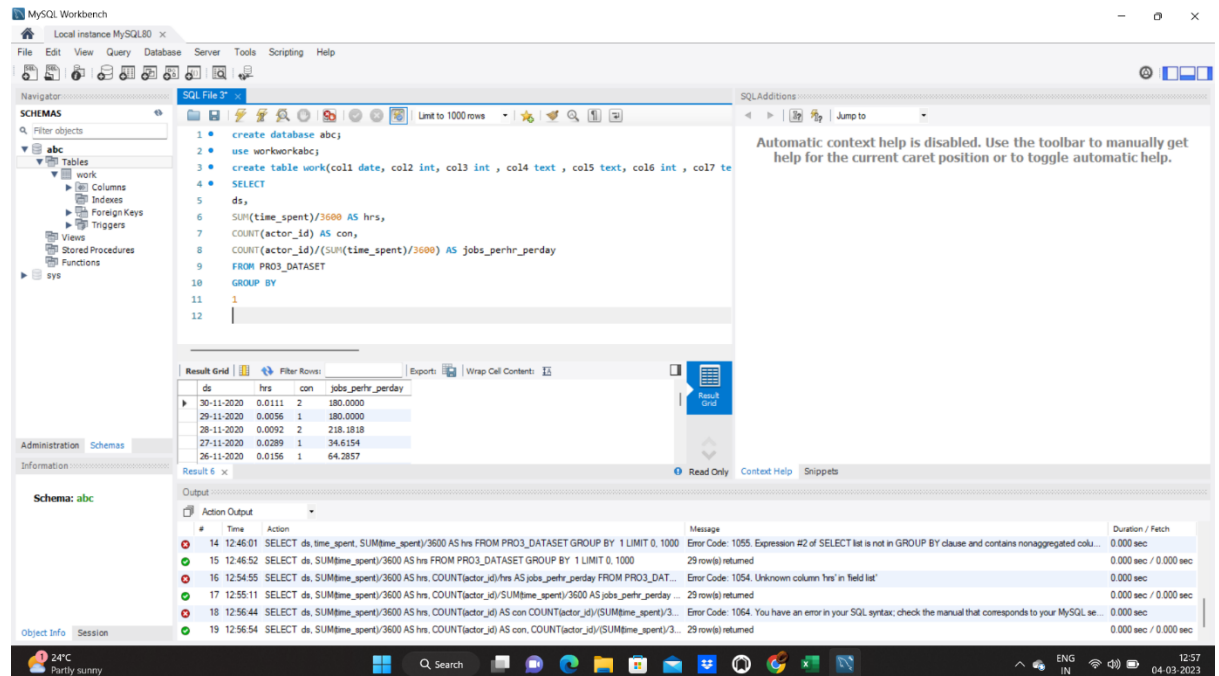
- MySQL Workbench 8.0 CE
- Excel

Insights:

Case Study 1 (Job Data)

- A. **Number of jobs reviewed:** Number of jobs reviewed over time.
Your task: Calculate the number of jobs reviewed per hour per day for November 2020?

Query:



The screenshot shows the MySQL Workbench interface. The SQL editor contains the following query:

```

1 create database abc;
2 use workabc;
3 create table work(col1 date, col2 int, col3 int , col4 text , col5 text, col6 int , col7 te
4 SELECT
5 ds,
6 SUM(time_spent)/3600 AS hrs,
7 COUNT(actor_id) AS con,
8 COUNT(actor_id)/(SUM(time_spent)/3600) AS jobs_perhr_perday
9 FROM PRO3_DATASET
10 GROUP BY
11 1
12

```

The Results grid shows the following data:

ds	hrs	con	jobs_perhr_perday
30-11-2020	0.0111	2	180.0000
29-11-2020	0.0056	1	180.0000
28-11-2020	0.0092	2	218.1818
27-11-2020	0.0289	1	34.6154
26-11-2020	0.0156	1	64.2857

The Output tab shows the execution log with the following messages:

#	Time	Action	Message	Duration / Fetch
14	12:46:01	SELECT ds, time_spent, SUM(time_spent)/3600 AS hrs FROM PRO3_DATASET GROUP BY 1 LIMIT 0, 1000	Error Code: 1055. Expression #2 of SELECT list is not in GROUP BY clause and contains nonaggregated colu...	0.000 sec
15	12:46:52	SELECT ds, SUM(time_spent)/3600 AS hrs FROM PRO3_DATASET GROUP BY 1 LIMIT 0, 1000	29 row(s) returned	0.000 sec / 0.000 sec
16	12:54:55	SELECT ds, SUM(time_spent)/3600 AS hrs, COUNT(actor_id) AS con, COUNT(actor_id)/SUM(time_spent)/3600 AS jobs_perhr_perday FROM PRO3_DAT...	Error Code: 1054. Unknown column 'hrs' in 'field list'	0.000 sec
17	12:55:11	SELECT ds, SUM(time_spent)/3600 AS hrs, COUNT(actor_id)/SUM(time_spent)/3600 AS jobs_perhr_perday ...	29 row(s) returned	0.000 sec / 0.000 sec
18	12:56:44	SELECT ds, SUM(time_spent)/3600 AS hrs, COUNT(actor_id) AS con, COUNT(actor_id)/SUM(time_spent)/3...	Error Code: 1064. You have an error in your SQL syntax; check the manual that corresponds to your MySQL se...	0.000 sec
19	12:56:54	SELECT ds, SUM(time_spent)/3600 AS hrs, COUNT(actor_id) AS con, COUNT(actor_id)/SUM(time_spent)/3...	29 row(s) returned	0.000 sec / 0.000 sec

According to the results, the number of jobs reviewed per hour per day is very high since it took only certain seconds to review a single job and not many jobs were reviewed on a single date i.e. the rate is high but not the work.

- B. **Throughput:** It is the no. of events happening per second.
Your task: Let's say the above metric is called throughput. Calculate 7 day rolling average of throughput? For throughput, do you prefer daily metric or 7-day rolling and why?

Query:

```

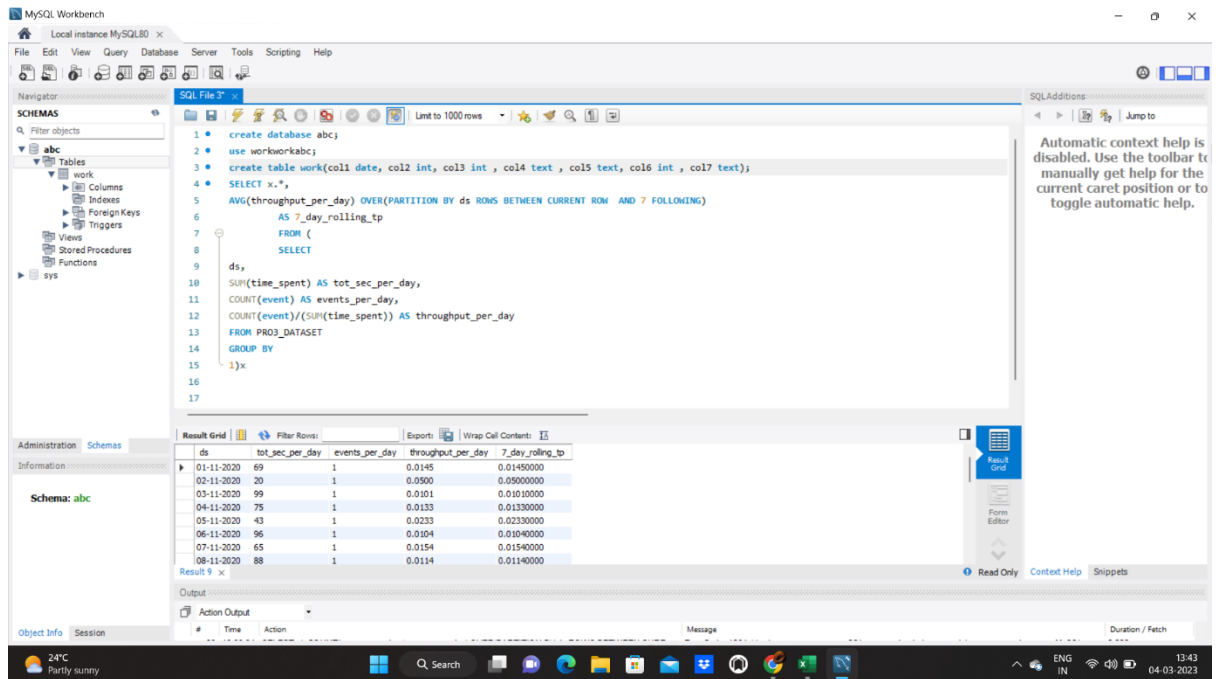
SELECT x.*,
AVG(throughput_per_day) OVER(PARTITION BY ds ROWS
BETWEEN CURRENT ROW AND 7 FOLLOWING)
AS 7_day_rolling_tp
FROM (
SELECT
ds,
SUM(time_spent) AS tot_sec_per_day,

```

```

COUNT(event) AS events_per_day,
COUNT(event)/(SUM(time_spent)) AS throughput_per_day
FROM PRO3_DATASET
GROUP BY
1)x

```



The table shows both the daily throughput and the 7 day rolling throughput. Not saying for this particular data in hand because both are quite small and doesn't make that much difference but 7 day rolling throughput is better than the daily because calculating and analysing rates on a daily basis doesn't makes sense not much difference occurs in a single day. It is better to derive insights on a substantial amount of data.

C. Percentage share of each language: Share of each language for different contents.

Your task: Calculate the percentage share of each language in the last 30 days?

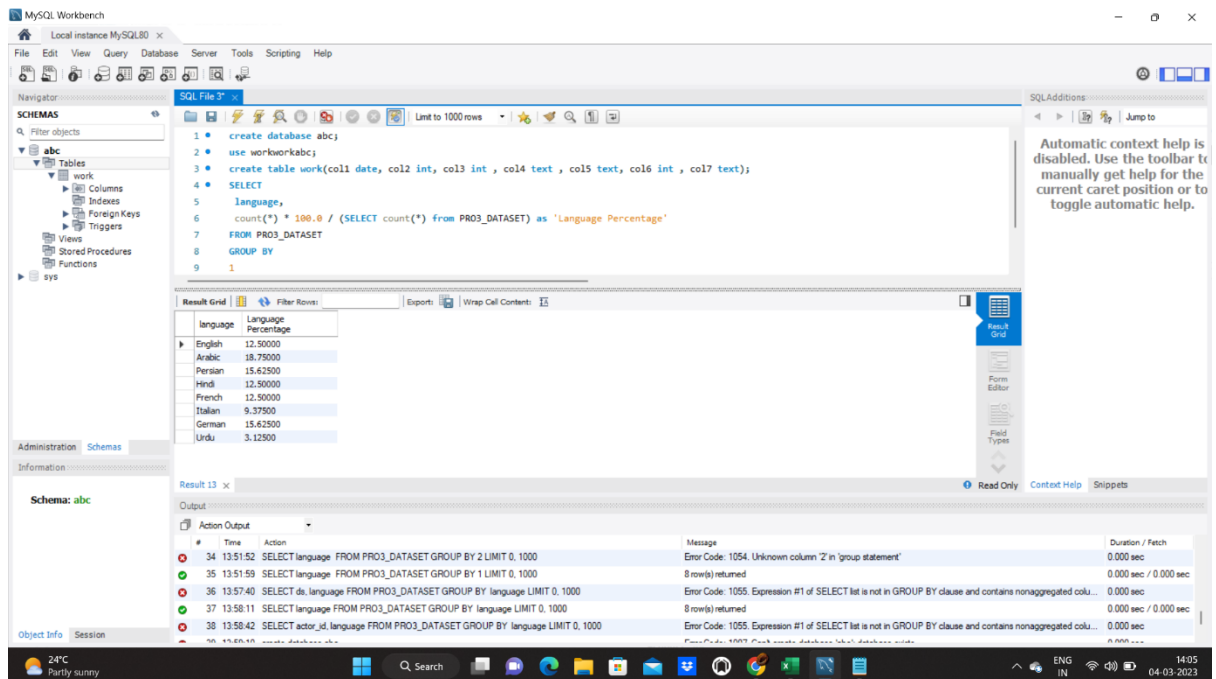
Query:

```

SELECT
language,
count(*) * 100.0 / (SELECT count(*) from PRO3_DATASET) as
'Language Percentage'
FROM PRO3_DATASET
GROUP BY

```

1



The results shows each language and the percentage of each language used in the month of November. The result is pretty straight forward.

D. Duplicate rows: Rows that have the same value present in them.

Your task: Let's say you see some duplicate rows in the data. How will you display duplicates from the table?

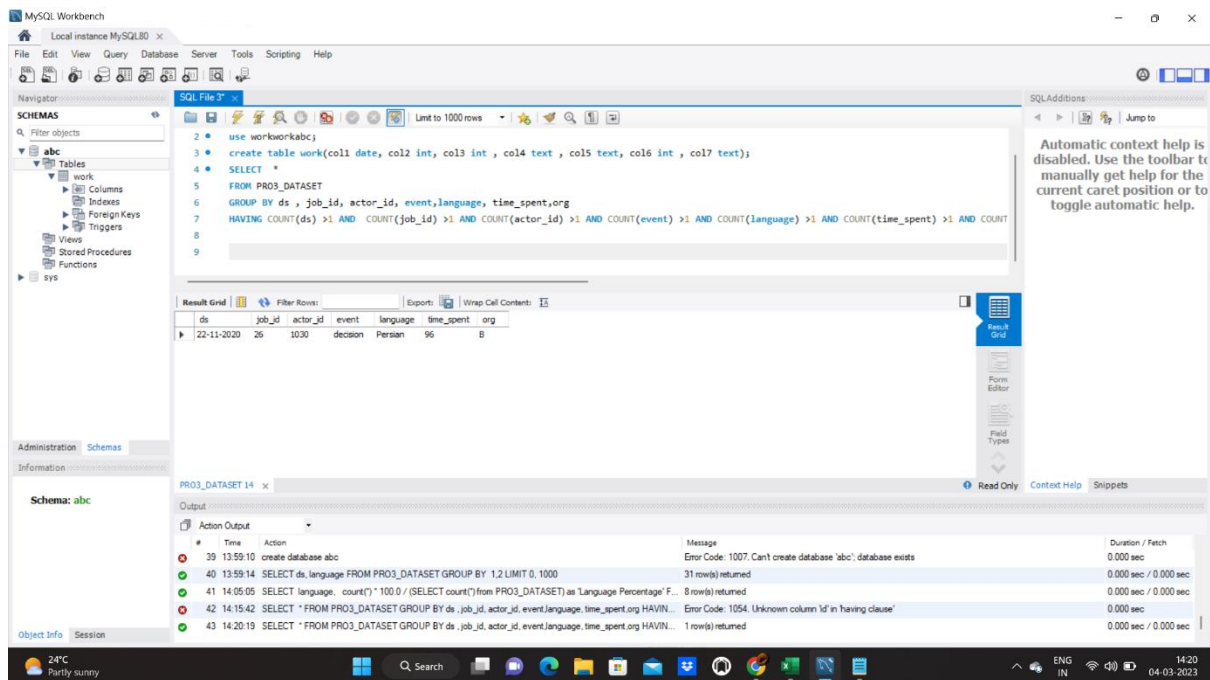
Query:

SELECT *

FROM PRO3_DATASET

GROUP BY ds , job_id, actor_id, event,language, time_spent,org

HAVING COUNT(ds) >1 AND COUNT(job_id) >1 AND COUNT(actor_id) >1 AND COUNT(event) >1 AND COUNT(language) >1 AND COUNT(time_spent) >1 AND COUNT(org) >1



The result showed there were two rows present in the dataset that were duplicates of each other that is both had every value of every column as the same.

Case Study 2 (Investigating metric spike)

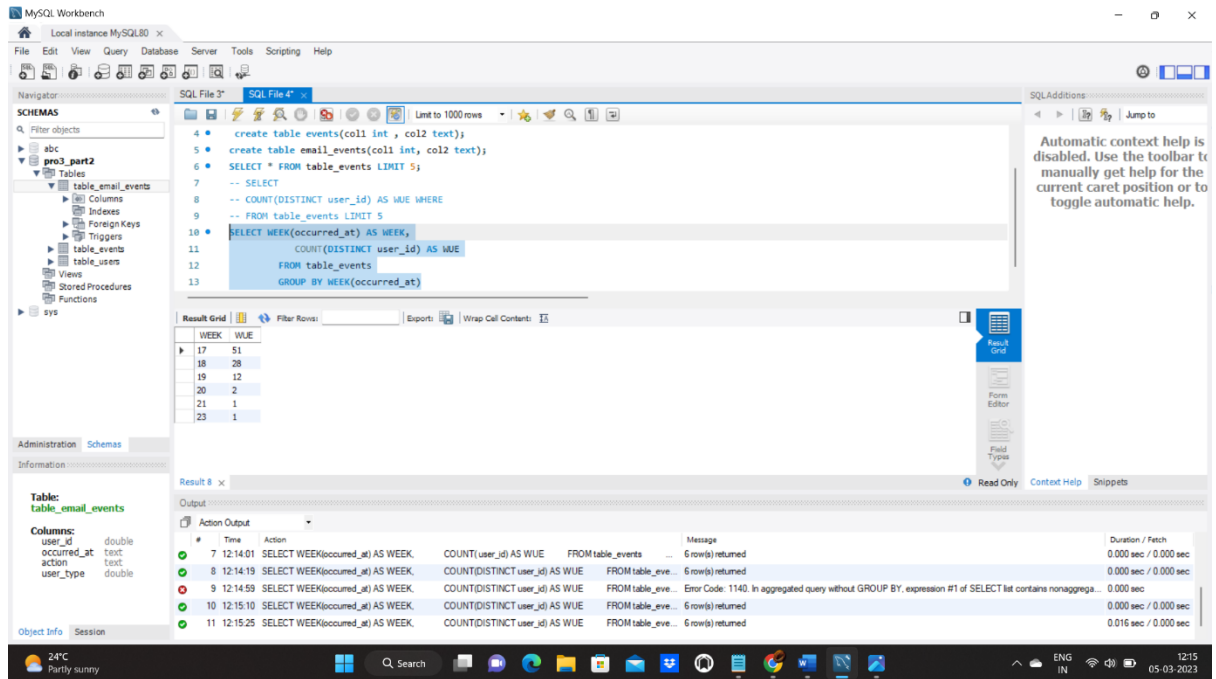
- A. **User Engagement:** To measure the activeness of a user. Measuring if the user finds quality in a product/service.
Your task: Calculate the weekly user engagement?

Query:

```

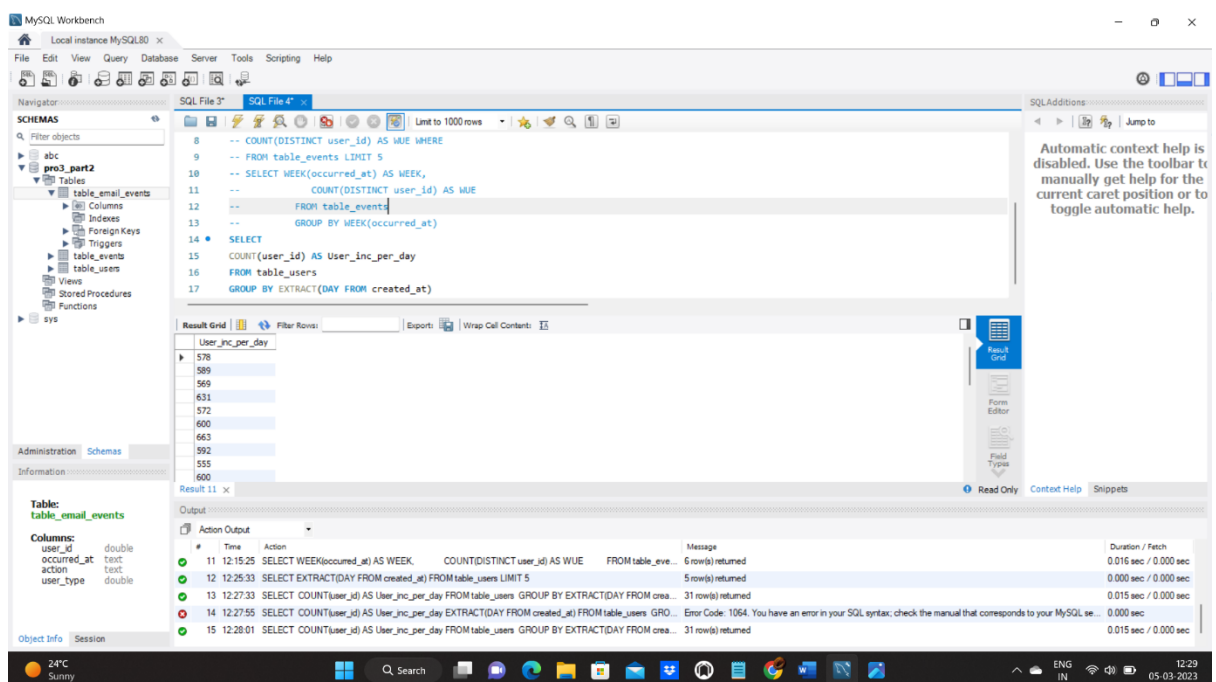
SELECT WEEK(occurred_at) AS WEEK,
        COUNT(DISTINCT user_id) AS WUE
FROM table_events
GROUP BY WEEK(occurred_at)

```



B. User Growth: Amount of users growing over time for a product.
Your task: Calculate the user growth for product?

Query:
SELECT
COUNT(user_id) AS User_inc_per_day
FROM table_users
GROUP BY EXTRACT(DAY FROM created_at)



The result shows the number of users that joined the product on the daily basis.

c. Weekly Retention: Users getting retained weekly after signing-up for a product.

Your task: Calculate the weekly retention of users-sign up cohort?

Query:

Select m.user_id,m.login_week,n.first as first,

m.login_week-first as week_number from

(SELECT user_id, EXTRACT(WEEK FROM occurred_at)

AS login_week FROM table_events WHERE event_name = "login" GROUP BY user_id ,

EXTRACT(WEEK FROM occurred_at)) m, (SELECT user_id,

MIN(EXTRACT(WEEK FROM occurred_at)) AS first FROM table_events WHERE event_name = "login"

The screenshot shows the MySQL Workbench interface. The SQL editor contains the following query:

```
-- FROM table_users
-- GROUP BY EXTRACT(DAY FROM created_at)

-- SELECT * FROM table_events WHERE event_name = "login"
Select m.user_id,m.login_week,n.first as first,
m.login_week-first as week_number from
(SELECT user_id, EXTRACT(WEEK FROM occurred_at)
AS login_week FROM table_events WHERE event_name = "login" GROUP BY user_id ,
EXTRACT(WEEK FROM occurred_at)) m, (SELECT user_id,
MIN(EXTRACT(WEEK FROM occurred_at)) AS first FROM table_events WHERE event_name = "login"
```

The Results Grid shows the following data:

user_id	login_week	first	week_number
10522	17	17	0
10612	17	17	0
10736	18	18	0
10965	19	19	0
11020	18	18	0
11037	17	17	0
11040	19	19	0
11133	18	18	0
11194	18	18	0
11212	18	18	0

The Output pane shows the execution of the query:

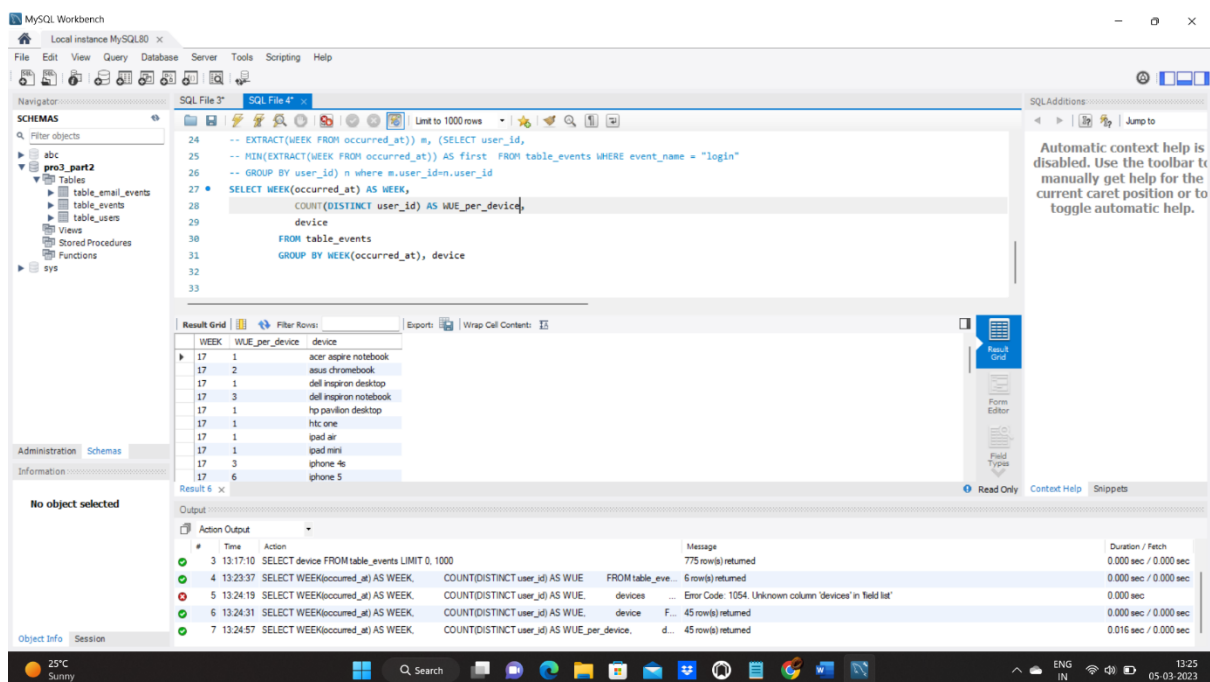
```
1 13:05:33 Select m.user_id,m.login_week,n.first as first, m.login_week-first as week_number from (SELECT user_id, DAT... 83 row(s) returned 0.000 sec / 0.000 sec
2 13:07:40 Select m.user_id,m.login_week,n.first as first, m.login_week-first as week_number from (SELECT user_id, EXT... 95 row(s) returned 0.000 sec / 0.000 sec
```


The results show excellent user retention and that most of the users returned to login in the same week with a very exceptions that returned the next week.

Weekly Engagement: To measure the activeness of a user. Measuring if the user finds quality in a product/service weekly.

Query:

```
SELECT WEEK(occurred_at) AS WEEK,  
  
       COUNT(DISTINCT user_id) AS WUE_per_device,  
  
       device  
  
FROM table_events  
  
GROUP BY WEEK(occurred_at), device
```



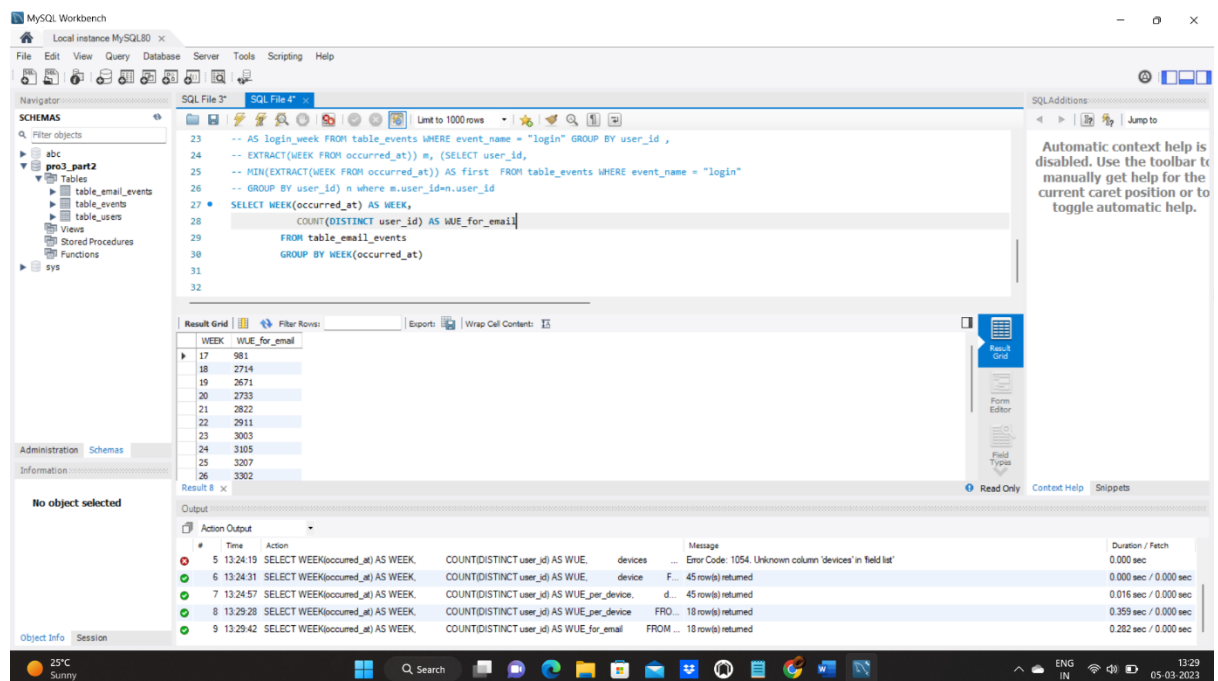
A. Email Engagement: Users engaging with the email service.

Your task: Calculate the email engagement metrics?

Query:

Calculating weekly user engagement for email services

```
SELECT WEEK(occurred_at) AS WEEK,  
  
        COUNT(DISTINCT user_id) AS WUE_for_email  
  
FROM table_email_events  
  
GROUP BY WEEK(occurred_at)
```



We now are calculating email service retention Select
m.user_id,m.login_week,n.first as first,

m.login_week-first as week_number from

(SELECT user_id, EXTRACT(WEEK FROM occurred_at)

AS login_week FROM table_email_events WHERE action =
"email_open" GROUP BY user_id ,

EXTRACT(WEEK FROM occurred_at)) m, (SELECT user_id,

MIN(EXTRACT(WEEK FROM occurred_at)) AS first FROM
table_email_events WHERE action = "email_open"

GROUP BY user_id) n where m.user_id=n.user_id

in users.

The screenshot shows the MySQL Workbench interface. The SQL editor contains a query that joins two tables, `table_email_events` and `table_email_events`, to calculate login weeks and first login times. The query is as follows:

```
24 -- EXTRACT(WEEK FROM occurred_at)) m, (SELECT user_id,
25 -- MIN(EXTRACT(WEEK FROM occurred_at)) AS first FROM table_email_events WHERE event_name = "login"
26 -- GROUP BY user_id) n where m.user_id=n.user_id
27
28 Select m.user_id,m.login_week,n.first as first,
29 m.login_week-first as week_number from
30 (SELECT user_id, EXTRACT(WEEK FROM occurred_at)
31 AS login_week FROM table_email_events WHERE action = "email_open" GROUP BY user_id ,
32 EXTRACT(WEEK FROM occurred_at)) m, (SELECT user_id,
33 MIN(EXTRACT(WEEK FROM occurred_at)) AS first FROM table_email_events WHERE action = "email_open"
```

The Results Grid shows the following data:

user_id	login_week	first	week_number
0	22	22	0
0	23	22	1
0	24	22	2
0	30	22	8
0	33	22	11
4	19	19	0
4	20	19	1
4	25	19	6
4	26	19	7
4	32	19	13

The Output pane shows the execution log with the following messages:

- 7 13:24:57 SELECT WEEK(occurred_at) AS WEEK, COUNT(DISTINCT user_id) AS WUE_per_device, d... 45 row(s) returned 0.016 sec / 0.000 sec
- 8 13:29:28 SELECT WEEK(occurred_at) AS WEEK, COUNT(DISTINCT user_id) AS WUE_per_device, FRO... 18 row(s) returned 0.359 sec / 0.000 sec
- 9 13:29:42 SELECT WEEK(occurred_at) AS WEEK, COUNT(DISTINCT user_id) AS WUE_per_email, FRO... 18 row(s) returned 0.282 sec / 0.000 sec
- 10 13:32:34 Select m.user_id,m.login_week,n.first as first, m.login_week-first as week_number from (SELECT user_id, EX... Error Code: 1054. Unknown column 'event_name' in 'where clause' 0.000 sec
- 11 13:33:26 Select m.user_id,m.login_week,n.first as first, m.login_week-first as week_number from (SELECT user_id, EX... 1000 row(s) returned 0.297 sec / 0.000 sec

Results:

While making this project I faced many challenges and had to search and for answering many of them but because of this I now have a much clear understanding of operation metrics and how is it used for not only gaining but also retaining the users in a service.

