# **PROJECT 3**

# Operation Analytics and Investigating Metric Spike.

# **Project Description:**

The company provided with 2 Datasets . In dataset 1 we have been given job\_data and in dataset 2 it consists of users data ,email\_events data , events data .It is the real-time data where we need to find meaningful trends and insights from the data .We were provided with Operational Analytics. Operational Analytics is a crucial process that involves analyzing a company's end-to-end operations. This analysis helps identify areas for improvement within the company.

Operational analytics is about using real-time data for daily decisions. Relevant business information flows from many sources into tools that analyze that data and identify problems and opportunities. This actionable data is then used by teams to inform their decision-making. companies must keep up with rapid changes in the business landscape. Operational analytics makes insights available in near-real-time for staff who need to address support tickets, repair remote equipment, or adjust pricing or sales tactics.

Operational analytics focuses on immediate action.

# <u>Keys</u>

- Operational analytics enables companies to keep pace with rapid changes in the business landscape.
- Data from multiple sources flows into systems that provide actionable insights on the fly.
- The benefits of operational analytics include smoother operations, lower costs, better products and services and ultimately happier customers.
- You can use operational analytics platforms in many fields and functions, such as agile development, customer support and predictive maintenance.

# Approach:

- -We were assigned with 2 Datasets. One dataset consist of Job\_Data.csv file and another dataset consist of user.csv ,events.csv and email event.csv files.
- I first filtered the data, and formatted the date time which was there in the file, then created a database and then imported the file in MYSQL WORKBENCH 8.0.
- I worked on each task /question provided and analysed basically on it , Then formatted and analyszed all the data and insights which are provided below.
- -It made some challenges with the questions to understand thoroughly, as it contains huge data in data set 2.
- -It was most challenging in figuring out the logics used in it and also the accurate solutions for it .
- -It helped me to work on real-time data .
- -This was the very first time I was working with huge dataset and analysing all the possible outcomes and then working on the solutions accurately.

# **Tech Stack:**

MYSQL WORKBENCH 8.0 CE

# **INSIGHTS:**

### Case Study 1: Job Data Analysis

### Tasks:

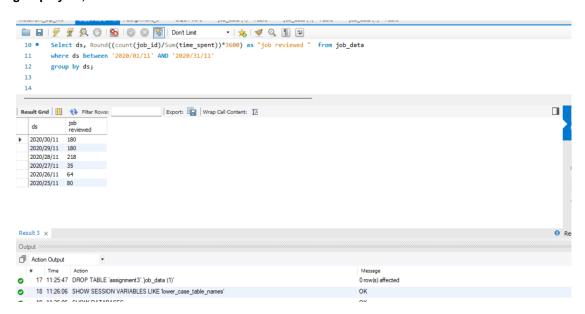
# A. Jobs Reviewed Over Time:

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

Answer: Select ds, Round((count(job\_id)/Sum(time\_spent))\*3600) as "job reviewed " from job\_data

where ds between '2020/01/11' AND '2020/31/11'

# group by ds;



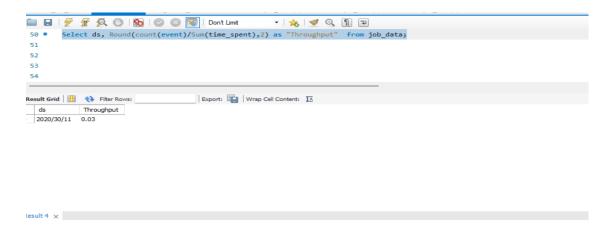
As a result, we can see that on 28th November 2020, the job review was the highest i.e 218.

В.

# **Throughput Analysis:**

- Objective: Calculate the 7-day rolling average of throughput (number of events per second).
- Your 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.

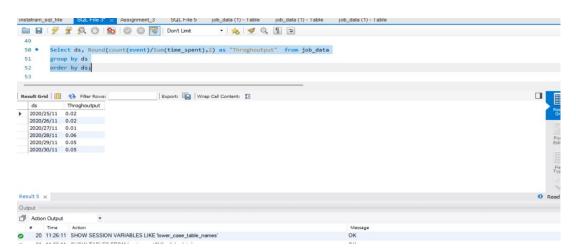
Answer: Select ds, Round(count(event)/Sum(time\_spent),2) as "Throughput" from job\_data;



It is a weekly throughoutput.

Select ds, Round(count(event)/Sum(time\_spent),2) as "Throghoutput" from job\_data group by ds

order by ds;



The maximum throughoutput is 0.06.

Metrics will go up and down on daily basis or a weekly basis. So, we will get many numbers faster everyday or every minute. Hence rolling metric will be better for this type of count because it calculates trends over short periods of time using a set of data.

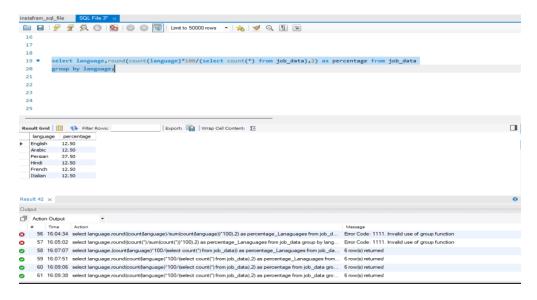
# C. Language Share Analysis:

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

Answer: select language,round(count(language)\*100/(select count(\*) from job\_data),2) as percentage

from job\_data

group by language;



Persian Language is the highest ,it is 37%, rounded 38%.

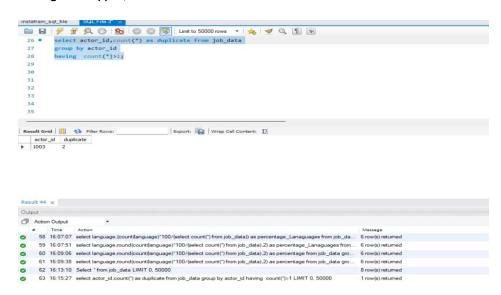
### D. **Duplicate Rows Detection:**

- Objective: Identify duplicate rows in the data.
- Your Task: Write an SQL query to display duplicate rows from the job\_data table.

Answer: select actor\_id,count(\*) as duplicate from job\_data

group by actor\_id

having count(\*)>1;



There are 2 duplicates for the ACTOR\_Id= 1003.

# Case Study 2: Investigating Metric Spike

# 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.

Answer: Select Extract(WEEK from occurred\_at) as "Week\_Number", Count( Distinct user\_id) as "User\_id" from events

where event\_type="engagement"

group by Week\_number;



### 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.

Answer: Select

months,Num\_of\_Activated\_Users,Round(((Num\_of\_Activated\_Users/LAG(Num\_of\_Activated\_Users,1) OVER (ORDER BY MONTHS)-1)\*100),2) as "growth in %"

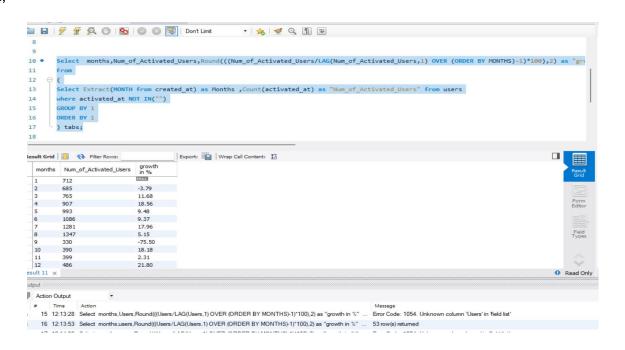
### From

(Select Extract(MONTH from created\_at) as Months ,Count(activated\_at) as "Num\_of\_Activated\_Users" from users where activated\_at NOT IN("")

### **GROUP BY**

## ORDER BY 1)

tabs;



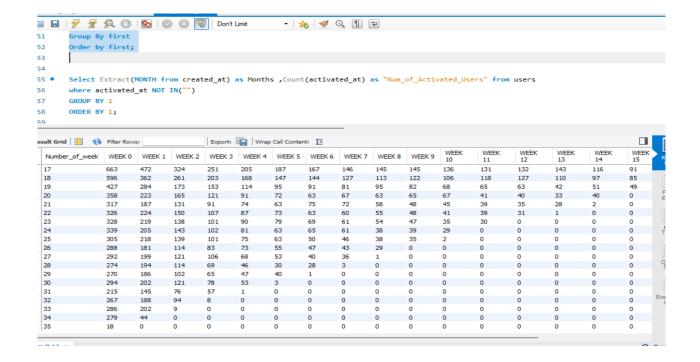
### C. Weekly Retention Analysis:

- 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.

```
SUM(CASE WHEN week_number=0 Then 1 else 0 END)AS "WEEK 0",
SUM(CASE WHEN week_number=1 Then 1 else 0 END)AS "WEEK 1",
SUM(CASE WHEN week_number=2 Then 1 else 0 END)AS "WEEK 2",
SUM(CASE WHEN week_number=3 Then 1 else 0 END)AS "WEEK 3",
SUM(CASE WHEN week_number=4 Then 1 else 0 END)AS "WEEK 4",
SUM(CASE WHEN week_number=5 Then 1 else 0 END)AS "WEEK 5",
SUM(CASE WHEN week_number=6 Then 1 else 0 END)AS "WEEK 6",
SUM(CASE WHEN week_number=7 Then 1 else 0 END)AS "WEEK 7",
SUM(CASE WHEN week_number=8 Then 1 else 0 END)AS "WEEK 8",
SUM(CASE WHEN week_number=9 Then 1 else 0 END)AS "WEEK 9",
SUM(CASE WHEN week_number=10 Then 1 else 0 END)AS "WEEK 10",
SUM(CASE WHEN week_number=11 Then 1 else 0 END)AS "WEEK 11",
SUM(CASE WHEN week_number=12 Then 1 else 0 END)AS "WEEK 12",
SUM(CASE WHEN week_number=13 Then 1 else 0 END)AS "WEEK 13",
SUM(CASE WHEN week_number=14 Then 1 else 0 END)AS "WEEK 14",
SUM(CASE WHEN week_number=15 Then 1 else 0 END)AS "WEEK 15",
SUM(CASE WHEN week_number=16 Then 1 else 0 END)AS "WEEK 16",
SUM(CASE WHEN week_number=17 Then 1 else 0 END)AS "WEEK 17",
SUM(CASE WHEN week_number=18 Then 1 else 0 END)AS "WEEK 18"
FROM
(
Select event1.user_id,event1.login_week,event2.first,event1.login_week-first AS Week_number
FROM
(
Select user_id, EXTRACT(WEEK FROM occurred_at) AS login_week from events
Group by user_id,login_week)event1,
(Select user_id, MIN(EXTRACT(WEEK FROM occurred_at)) AS First from events
Group by user_id)event2
Where event1.user_id=event2.user_id
)tabs
Group By first
```

Order by first;

Answer: Select first as "Number\_of\_week",



### D. Weekly Engagement Per Device:

- Objective: Measure the activeness of users on a weekly basis per device.
- o Your Task: Write an SQL query to calculate the weekly engagement per device.

Answer: Select EXTRACT(WEEK from occurred\_at) As "WEEK NUMBERS",

Count(Distinct CASE WHEN device IN('dell inspiron notebook') Then user\_id ELSE NULL END)AS

"Dell inspiron notebook",

Count(Distinct CASE WHEN device IN('iphone 5') Then user\_id ELSE NULL END)AS

"iPhone 5",

Count(Distinct CASE WHEN device IN('iphone 4s') Then user\_id ELSE NULL END)AS

"iPhone 4s",

Count(Distinct CASE WHEN device IN('windows surface') Then user\_id ELSE NULL END)AS

"Windows surface",

Count(Distinct CASE WHEN device IN('macbook air') Then user\_id ELSE NULL END)AS

"Macbook air",

Count(Distinct CASE WHEN device IN('iphone 5s') Then user\_id ELSE NULL END)AS

"iPhone 5s",

Count(Distinct CASE WHEN device IN('macbook pro') Then user\_id ELSE NULL END)AS

"Macbook pro",

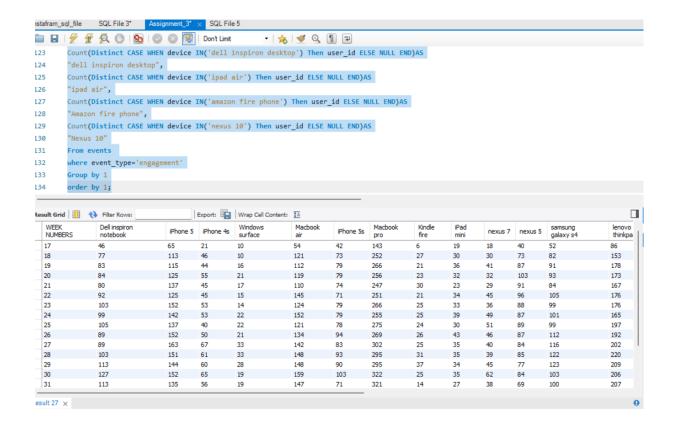
Count(Distinct CASE WHEN device IN('kindle fire') Then user\_id ELSE NULL END)AS

"Kindle fire",

Count(Distinct CASE WHEN device IN('ipad mini') Then user\_id ELSE NULL END)AS

"iPad mini",

```
Count(Distinct CASE WHEN device IN('nexus 7') Then user_id ELSE NULL END)AS
"nexus 7",
Count(Distinct CASE WHEN device IN('nexus 5') Then user_id ELSE NULL END)AS
"nexus 5",
Count(Distinct CASE WHEN device IN('samsung galaxy s4') Then user_id ELSE NULL END)AS
"samsung galaxy s4",
Count(Distinct CASE WHEN device IN('lenovo thinkpad') Then user_id ELSE NULL END)AS
"lenovo thinkpad",
Count(Distinct CASE WHEN device IN('samsumg galaxy tablet') Then user_id ELSE NULL END)AS
"samsumg galaxy tablet",
Count(Distinct CASE WHEN device IN('asus chromebook') Then user_id ELSE NULL END)AS
"asus chromebook",
Count(Distinct CASE WHEN device IN('htc one') Then user_id ELSE NULL END)AS
"HTC one",
Count(Distinct CASE WHEN device IN('nokia lumia 635') Then user_id ELSE NULL END)AS
"Nokia lumia 635",
Count(Distinct CASE WHEN device IN('mac mini') Then user_id ELSE NULL END)AS
"mac mini",
Count(Distinct CASE WHEN device IN('dell inspiron desktop') Then user_id ELSE NULL END)AS
"dell inspiron desktop",
Count(Distinct CASE WHEN device IN('ipad air') Then user_id ELSE NULL END)AS
"ipad air",
Count(Distinct CASE WHEN device IN('amazon fire phone') Then user_id ELSE NULL END)AS
"Amazon fire phone",
Count(Distinct CASE WHEN device IN('nexus 10') Then user_id ELSE NULL END)AS
"Nexus 10"
From events
where event_type='engagement'
Group by 1
order by 1;
```



### 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.

Answer: SELECT WEEK,

ROUND((weekly\_digest/total\*100),2) AS "Weekly Digest Rate",

Round((email\_opens/total\*100),2) AS "Email\_open\_rate",

Round((email\_clickthroughs/total\*100),2) AS "Email Clickthrough Rate",

Round((reengagement\_emails/total\*100),2) AS "Reengagement Email Rate"

From

(

Select EXTRACT(WEEK from occurred\_at) AS WEEK,

Count(CASE WHEN action="sent\_weekly\_digest" THEN user\_id ELSE NULL END) AS weekly\_digest,

Count(CASE WHEN action="email\_open" THEN user\_id ELSE NULL END) AS email\_opens,

Count(CASE WHEN action="email\_clickthrough" THEN user\_id ELSE NULL END) As email\_clickthroughs,

Count(CASE WHEN action="sent\_reengagement\_email" THEN user\_id ELSE NULL END) As reengagement\_emails,

Count(user\_id) AS total

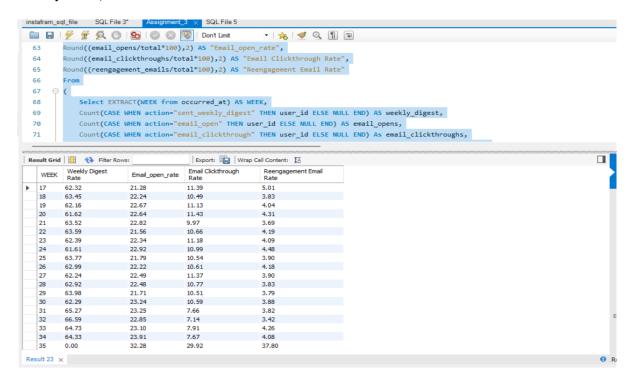
From email\_events

Group BY 1

)tabs

### Group By week

### order By WEEK;



# Results:

The project was the best way to boost up with my SQL knowledge. I have worked on throughout to analyse weekly and monthly data. I came to know about the data cleaning and data formatting in this. I can get the desired output while writing the queries.

It was the most difficult task as it consists to real time data and business insights. It helped me to brush up my knowledge on SQL and as well as, it helped me to face some challenges while working on Advanced SQL.

This Encouraged me to solve more problems until I am confident with it and helped me to create beautiful insights of the data.