# Operation Analytics and Investigating Metric Spike

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#### **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.
- Being one of the most important parts of a company, this kind of analysis is further used to predict the overall growth or decline of a company's fortune. It means better automation, better understanding between cross-functional teams, and more effective workflows.
- 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.
- > You are working for a company like Microsoft designated as Data Analyst Lead and is provided with different data sets, tables from which you must derive certain insights out of it and answer the questions asked by different departments.
- We need to provide a case study on the the given datasets.

#### **APPROACH:**

We need to import our dataset i.e; tables into mysql and using different sql operations the following questions and analysis must be done.

TECHSTACK USED: MYSQL WORKBENCH

#### Case Study 1 (Job Data)

Below is the structure of the table with the definition of each column that you must work on:

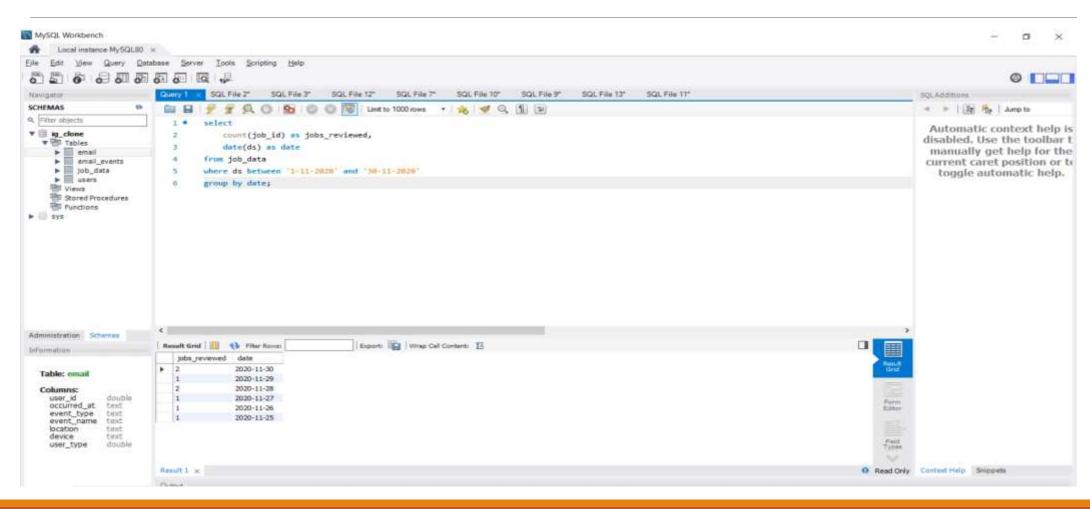
- ➤ Table-1: job\_data
- job\_id: unique identifier of jobs
- >actor\_id: unique identifier of actor
- > event: decision/skip/transfer
- ► language: language of the content
- time\_spent: time spent to review the job in seconds
- ▶org: organization of the actor
- >ds: date in the yyyy/mm/dd format. It is stored in the form of text and we use presto to run. no need for date function

#### Case Study 2 (Investigating metric spike)

The structure of the table with the definition of each column that you must work on is present in the project image

- ► Table-1: users
- >This table includes one row per user, with descriptive information about that user's account.
- ► Table-2: events
- This table includes one row per event, where an event is an action that a user has taken. These events include login events, messaging events, search events, events logged as users progress through a signup funnel, events around received emails.
- ➤ Table-3: email\_events
- This table contains events specific to the sending of emails. It is similar in structure to the events table above.

- Number of jobs reviewed: Amount of jobs reviewed over time.
- ➤ Your task: Calculate the number of jobs reviewed per hour per day for November 2020?
- >QUERY:
- > select
- >count(job\_id) as jobs\_reviewed,
- >date(ds) as date
- >from job\_data
- > where ds between '1-11-2020' and '30-11-2020'
- >group by date;



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?

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QUERY:

SELECT ds, jobs_reviewed,

AVG(jobs_reviewed)OVER(ORDER BY ds ROWS BETWEEN 6 PRECEDING AND

CURRENT ROW)AS 7_day_rolling_avg FROM

(

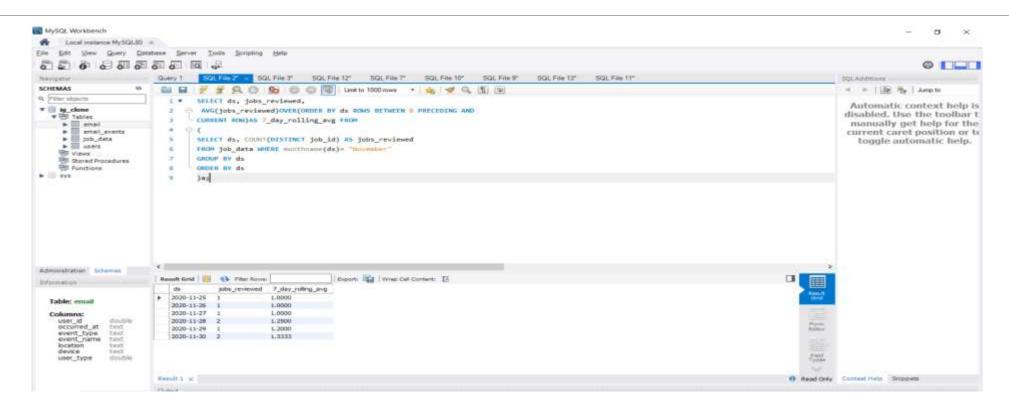
SELECT ds, COUNT(DISTINCT job_id) AS jobs_reviewed

FROM job_data WHERE monthname(ds)= "November"

GROUP BY ds

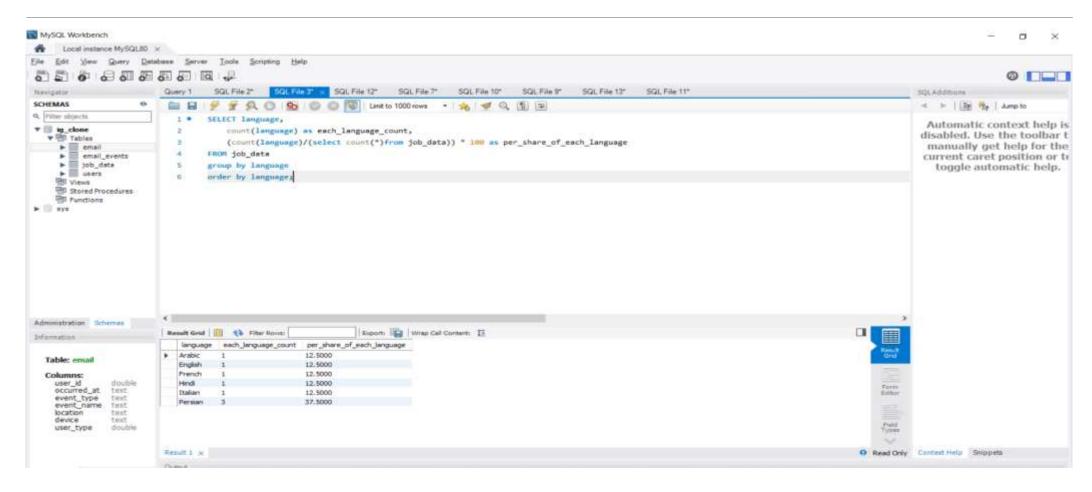
ORDER BY ds

)a;
```



output:7 day rolling is prefarable for detail aspects as in each day most the events are not happening.

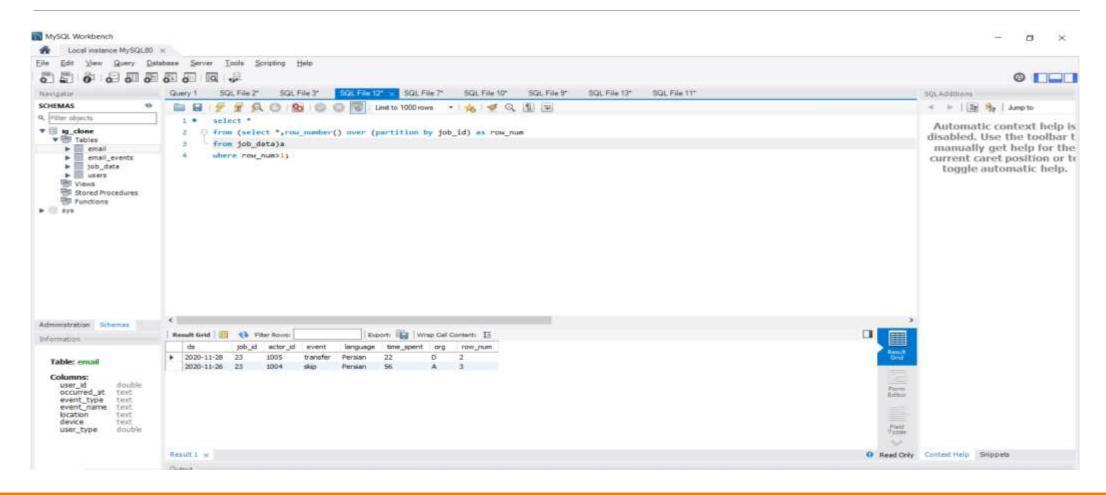
Percentage share of each language: Share of each language for different contents. Your task: Calculate the percentage share of each language in the last 30 days? **QUERY:** SELECT language, count(language) as each\_language\_count, (count(language)/(select count(\*)from job\_data)) \* 100 as per\_share\_of\_each\_language FROM job\_data group by language order by language;



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 (select *,row_number() over (partition by job_id) as row_num
from jobdata)a
where row_num>1;
```



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?

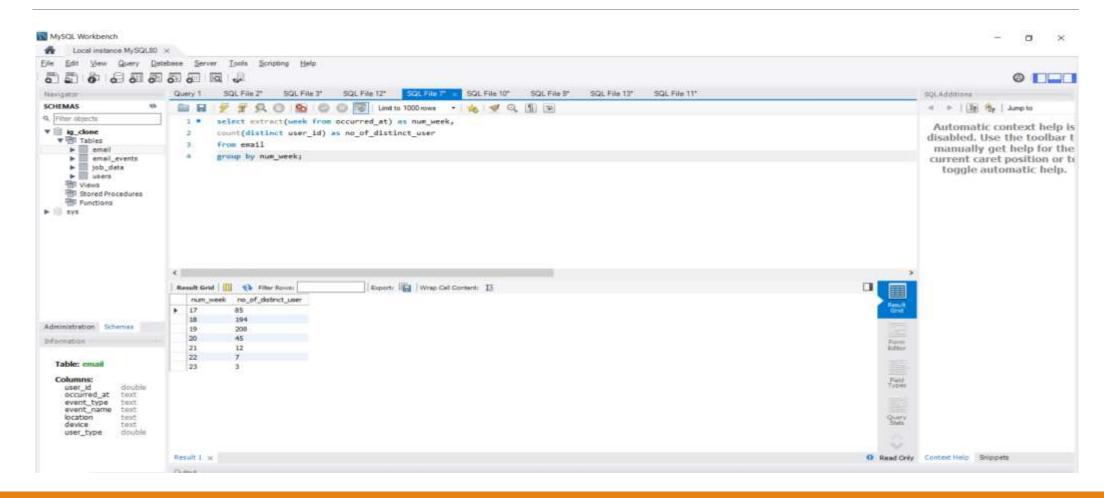
QUERY:

select extract(week from occurred\_at) as num\_week,

count(distinct user\_id) as no\_of\_distinct\_user

from email

group by num\_week;



User Growth: Amount of users growing over time for a product.

Your task: Calculate the user growth for product?

#### QUERY:

select weeknum, year, num\_active\_user,

SUM(num\_active\_user)over(order by year, weeknum rows between unbounded preceding and current row)as cum\_active\_users

**FROM** 

(select

extract(year from activated\_at) as year,

extract(week from activated\_at)as weeknum,

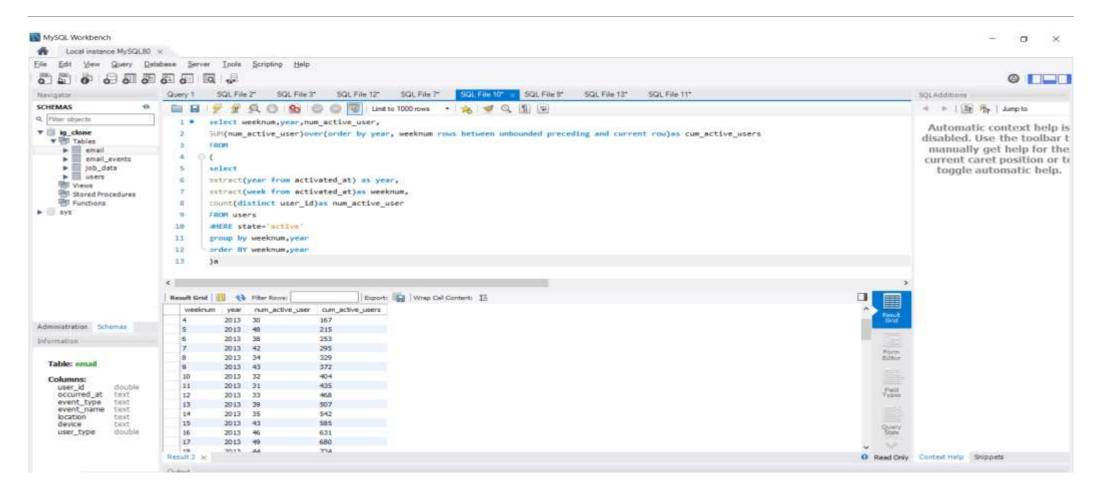
count(distinct user\_id)as num\_active\_user

FROM users

WHERE state='active'

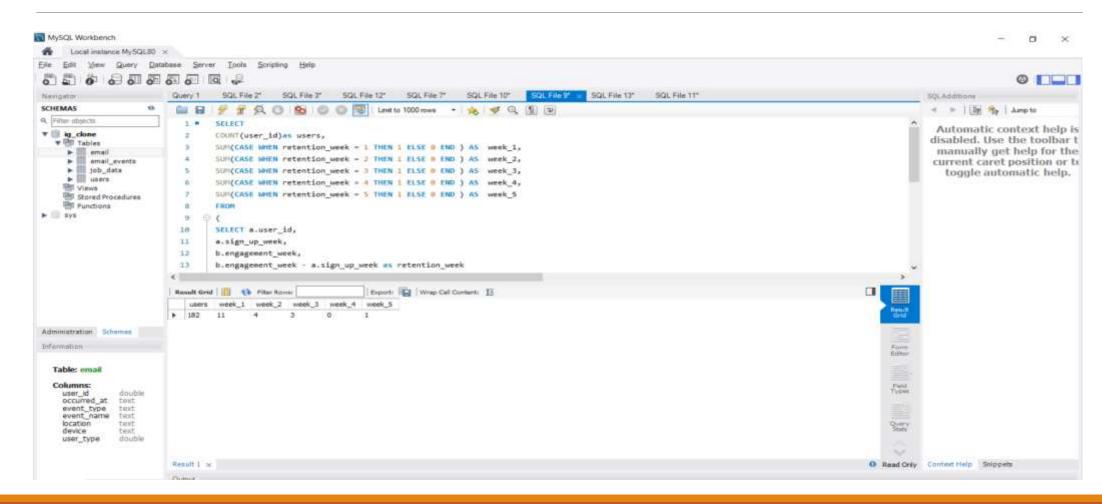
group by weeknum, year

order BY weeknum, year)a



```
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
COUNT(user id)as users,
SUM(CASE WHEN retention week = 1 THEN 1 ELSE 0 END ) AS week 1,
SUM(CASE WHEN retention week = 2 THEN 1 ELSE 0 END ) AS week 2,
SUM(CASE WHEN retention week = 3 THEN 1 ELSE 0 END ) AS week 3,
SUM(CASE WHEN retention week = 4 THEN 1 ELSE 0 END ) AS week 4,
SUM(CASE WHEN retention week = 5 THEN 1 ELSE 0 END ) AS week 5
FROM
```

```
FROM (SELECT a.user id,
a.sign_up_week,
b.engagement week,
b.engagement_week - a.sign_up_week as retention_week FROM (
(select distinct user id, extract(week from occurred at) as sign up week
from email
where event type='signup flow'
and event name='complete_signup'
and extract(week from occurred at)=18)a
left JOIN (
select distinct user id,
extract(week from occurred_at) as engagement_week
from email
where event type='engagement')b
on a.user_id=b.user_id )
order by
a.user id )a
```



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

Your task: Calculate the weekly engagement per device?

QUERY:

select

extract(year from occurred at)as year,

extract(week from occurred\_at)as week,

device,

COUNT(distinct user\_id) as num\_users

FROM

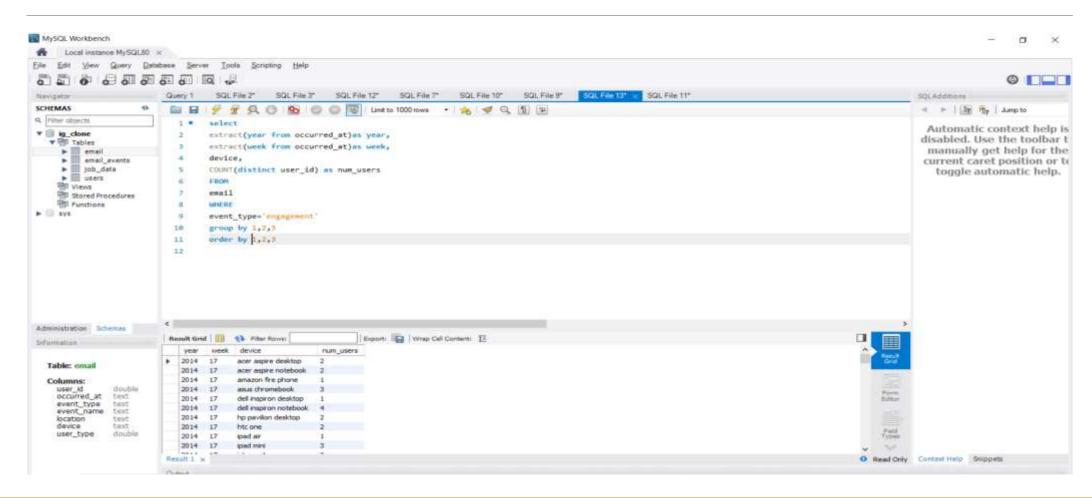
email

WHERE

event\_type='engagement'

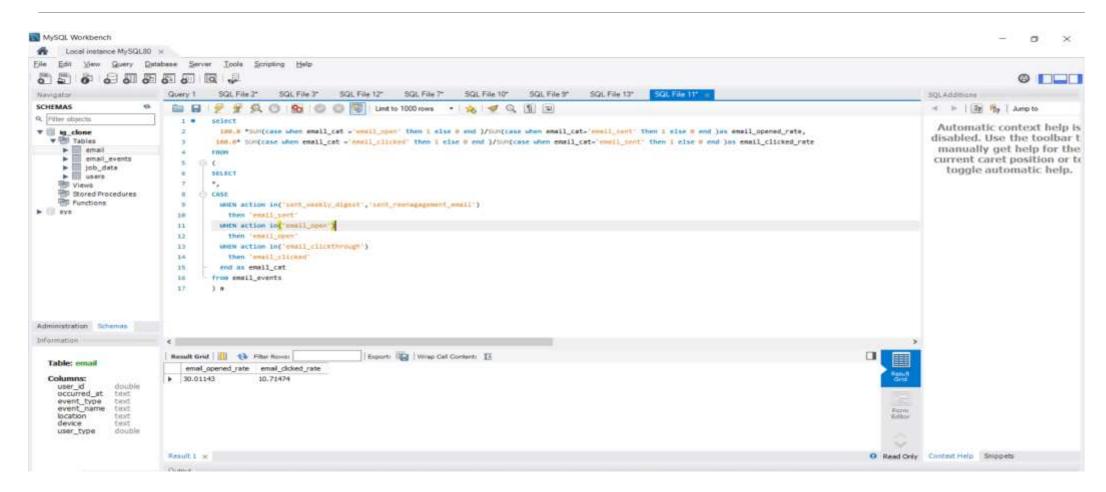
group by 1,2,3

order by 1,2,3



Email Engagement: Users engaging with the email service.

Your task: Calculate the email engagement metrics? QUERY: select 100.0 \*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 opened rate, 100.0\* 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 clicked rate FROM (SELECT \*,CASE WHEN action in ('sent weekly digest', 'sent reenagagement 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 ) a



### **RESULT**

We have learned different sql functions like :

- extract
- date
- avg
- over
- distinction
- and many more.