

Trainity project 3

Operation Analytics and Investigating Metric Spike

Project Description:

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. As a Data Analyst, you'll work closely with various teams, such as operations, support, and marketing, helping them derive valuable insights from the data they collect.

Approach:

First we extracted the data given into tables and analyzed user engagement, event counts, and distribution patterns, comparing them across time periods to pinpoint irregularities etc.

Tech-stacks used:

We extracted and cleaned the table data in excel and loaded the table data to mysql to write the sql queries to analyze the data.

Insights:

Identified a significant spike in event generation during specific weeks.

Observed that the average events per user and possible bot activity.

Weekly user engagement trends revealed periods of high variability correlating with specific marketing campaigns.

Result:

The analysis successfully pinpointed the source, enabling corrective measures to be taken. This project enhanced our understanding of operational metrics, user behavior, and how external factors influence engagement, driving more informed decision-making.

Case Study 1: Job Data Analysis

You will be working with a 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).

To create the table `job_data` from the given data

Initial code:

```
create table job_data
```

```
(ds date,
```

```
job_id int not null,
```

```
actor_id int not null,
```

```
event varchar(50) not null,
```

```
language varchar(50) not null,
```

```
time_spent int not null,
```

```
org char(2)
```

```
);
```

```
INSERT INTO job_data(ds, job_id, actor_id, event, language, time_spent, org)
```

```
VALUES
```

```
('2020-11-30', 21, 1001, 'skip', 'English', 15, 'A'),
```

```
('2020-11-30', 22, 1006, 'transfer', 'Arabic', 25, 'B'),
```

```
('2020-11-29', 23, 1003, 'decision', 'Persian', 20, 'C'),
```

```
('2020-11-28', 23, 1005, 'transfer', 'Persian', 22, 'D'),
```

```
('2020-11-28', 25, 1002, 'decision', 'Hindi', 11, 'B'),
```

```
('2020-11-27', 11, 1007, 'decision', 'French', 104, 'D'),
```

```
('2020-11-26', 23, 1004, 'skip', 'Persian', 56, 'A'),  
('2020-11-25', 20, 1003, 'transfer', 'Italian', 45, 'C');
```

Tasks:

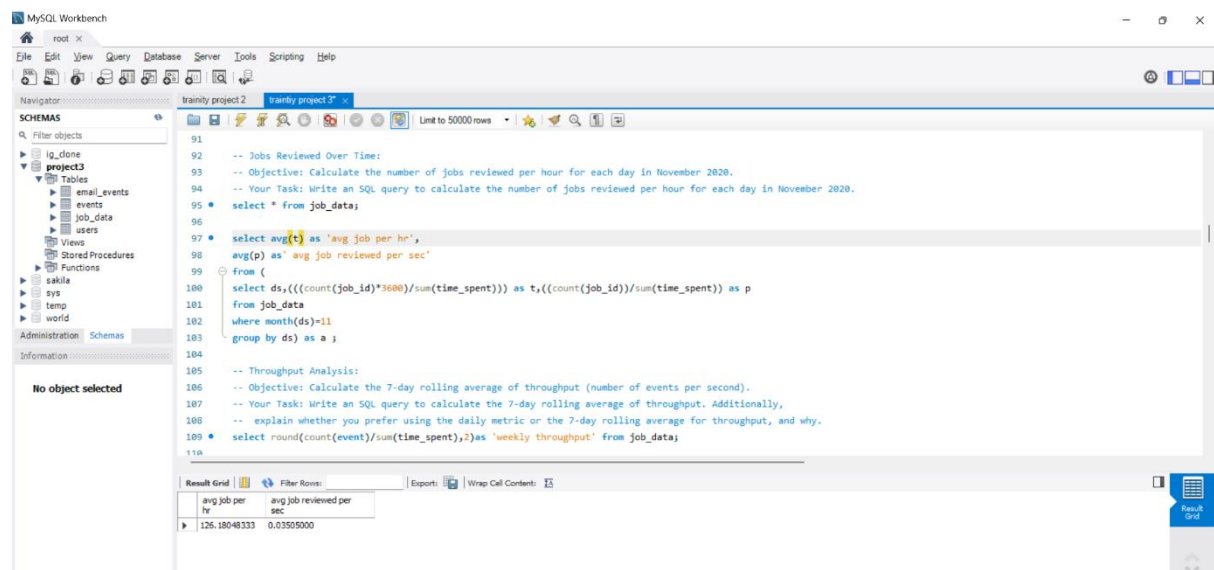
A. Jobs Reviewed Over Time:

- 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.

Code:

```
select avg(t) as 'avg job per hr',  
avg(p) as 'avg job reviewed per sec'  
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) as a ;
```

Output:



The screenshot shows the MySQL Workbench interface. The left sidebar displays the 'SCHEMAS' panel with a tree view of databases including 'project3'. The main editor window contains a SQL query for calculating average jobs reviewed per hour and per second for November 2020. The query is as follows:

```
-- 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.  
select * from job_data;  
  
select avg(t) as 'avg job per hr',  
avg(p) as 'avg job reviewed per sec'  
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) as a ;  
  
-- 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.  
select round(count(event)/sum(time_spent),2)as 'weekly throughput' from job_data;
```

At the bottom, the 'Result Grid' shows the output of the query:

avg job per hr	avg job reviewed per sec
126.18048333	0.03505000

B. Throughput Analysis:

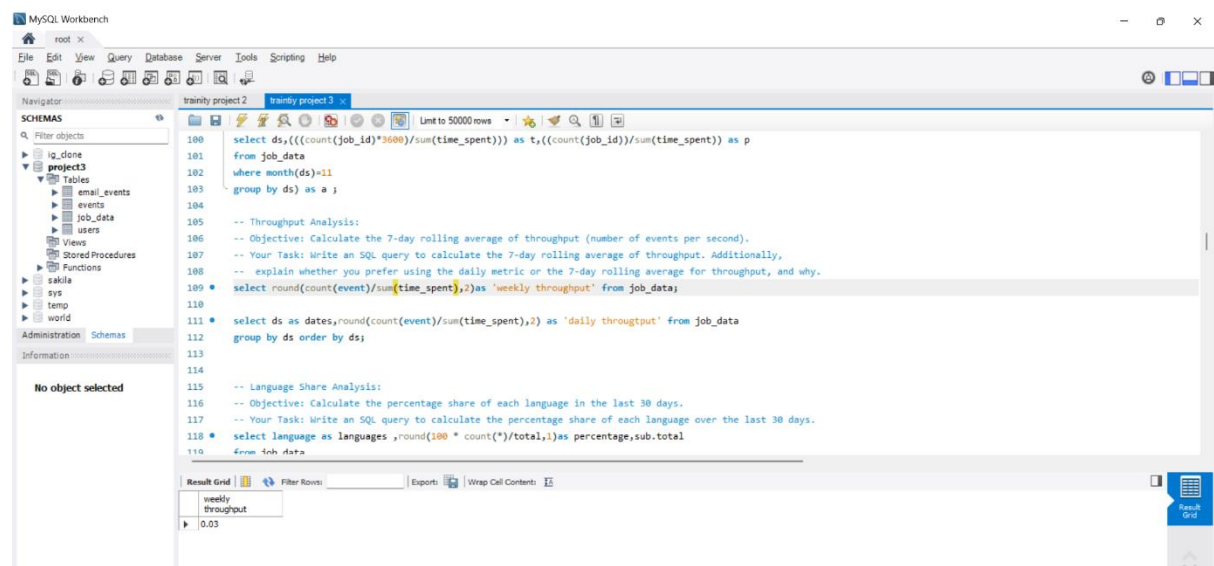
- Objective: Calculate the 7-day rolling average of throughput (number of events per second).
- Your Task: Write an SQL query to calculate the 7-day rolling average of throughput. Additionally, explain whether you prefer using the daily metric or the 7-day rolling average for throughput, and why.

Code:

```
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;
```

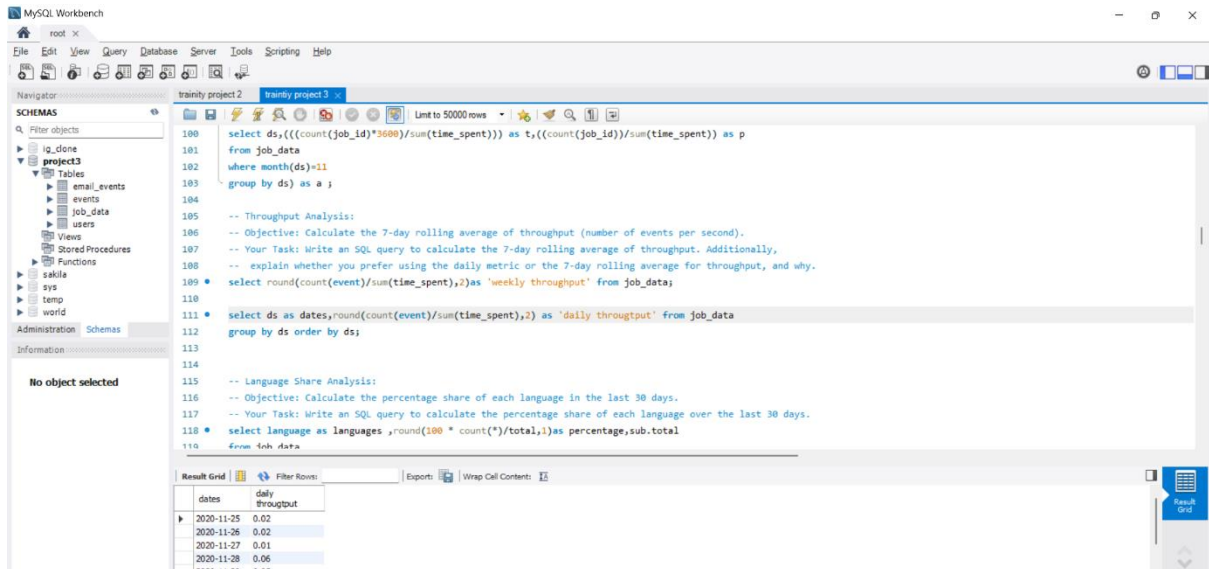
Output:



The screenshot shows the MySQL Workbench interface. The left sidebar displays the 'SCHEMAS' panel with a tree view of databases including 'project3' and 'job_data'. The main editor window contains a SQL query with line numbers 100 through 118. The query calculates the 7-day rolling average of throughput. The output pane at the bottom shows a single row with the column 'weekly throughput' and the value '0.03'.

```
100 select ds,(((count(job_id)*3600)/sum(time_spent))) as t,((count(job_id))/sum(time_spent)) as p  
101 from job_data  
102 where month(ds)=11  
103 group by ds) as a ;  
104  
105 -- Throughput Analysis:  
106 -- Objective: Calculate the 7-day rolling average of throughput (number of events per second).  
107 -- Your Task: Write an SQL query to calculate the 7-day rolling average of throughput. Additionally,  
108 -- explain whether you prefer using the daily metric or the 7-day rolling average for throughput, and why.  
109 • select round(count(event)/sum(time_spent),2)as 'weekly throughput' from job_data;  
110  
111 • select ds as dates,round(count(event)/sum(time_spent),2) as 'daily throughput' from job_data  
112 group by ds order by ds;  
113  
114  
115 -- Language Share Analysis:  
116 -- Objective: Calculate the percentage share of each language in the last 30 days.  
117 -- Your Task: Write an SQL query to calculate the percentage share of each language over the last 30 days.  
118 • select language as languages ,round(100 * count(*)/total,1)as percentage,sub.total  
from job_data
```

weekly throughput
0.03



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.

Code:

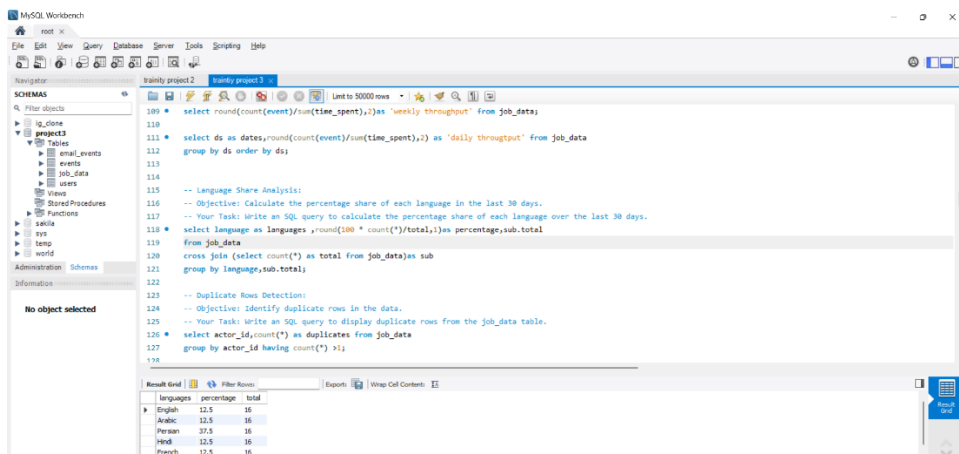
```
select language as languages ,round(100 * count(*)/total,1)as percentage,sub.total
```

```
from job_data
```

```
cross join (select count(*) as total from job_data)as sub
```

```
group by language,sub.total;
```

Output:



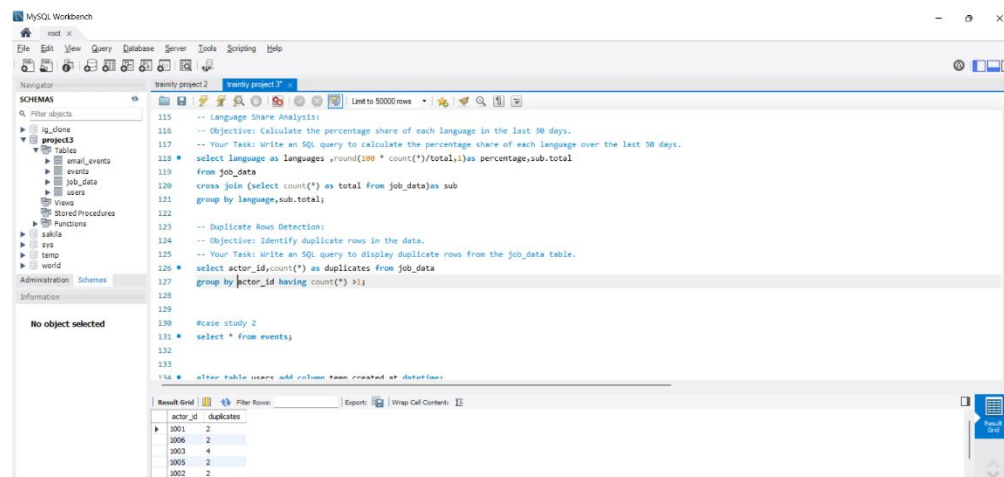
D. Duplicate Rows Detection:

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

Code:

```
select actor_id,count(*) as duplicates from job_data  
group by actor_id having count(*) >1;
```

Output:



Case Study 2: Investigating Metric Spike

You will be working with 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.

After cleaning the tables data in excel we have to load the tables in mysql

Initial code:

```
create table users(  
  user_id int,  
  created_at varchar(100),  
  company_id int,  
  language varchar(100),
```

```
activated_at varchar(100),  
state varchar(100));
```

```
show variables like 'secure_file_priv';
```

```
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;
```

```
select * from users;
```

```
create table events(  
user_id int,  
occurred_at varchar(100),  
event_type varchar(100),  
event_name varchar(100),  
location varchar(100),  
device varchar(100),  
user_type int);
```

```
load data infile "C:/ProgramData/MySQL/MySQL Server 8.0/Uploads/events.csv"  
into table events  
fields terminated by ','  
enclosed by '"'  
lines terminated by '\n'  
ignore 1 rows;
```

```
select * from events;
```

```
create table email_events(  
  user_id int,  
  occurred_at varchar(100),  
  action varchar(100),  
  user_type int);
```

```
load data infile "C:/ProgramData/MySQL/MySQL Server 8.0/Uploads/email_events.csv"  
into table email_events  
fields terminated by ','  
enclosed by '"'  
lines terminated by '\n'  
ignore 1 rows;
```

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.

Code:

```
WITH weekly_activity AS (  
  SELECT  
    user_id,  
    DATE_TRUNC('week', occurred_at) AS week_start,  
    COUNT(*) AS event_count  
  FROM  
    events  
  GROUP BY  
    user_id, DATE_TRUNC('week', occurred_at)
```


)

SELECT

week_start,

COUNT(DISTINCT user_id) AS active_users,

AVG(event_count) AS avg_events_per_user

FROM

weekly_activity

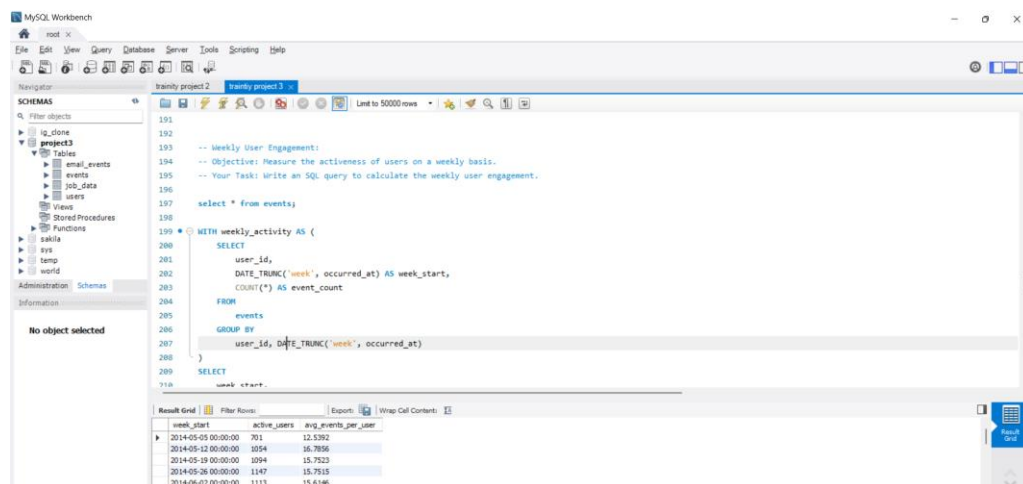
GROUP BY

week_start

ORDER BY

week_start;

Output:



The screenshot shows the MySQL Workbench interface. The SQL editor contains a query to calculate weekly user engagement. The query uses a CTE named 'weekly_activity' to calculate the average number of events per user for each week, based on the 'events' table. The main query then selects the week_start, the count of distinct active users, and the average events per user from the 'weekly_activity' CTE, grouped by week_start.

Result Grid:

week_start	active_users	avg_events_per_user
2014-05-05 00:00:00	701	12.5392
2014-05-12 00:00:00	1054	16.7856
2014-05-19 00:00:00	1094	15.7523
2014-05-26 00:00:00	1147	15.7515
2014-06-02 00:00:00	1113	15.6146

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.

Code:

SELECT

DATE_TRUNC('week', activated_at) AS week_start,

COUNT(*) AS new_users

FROM

users

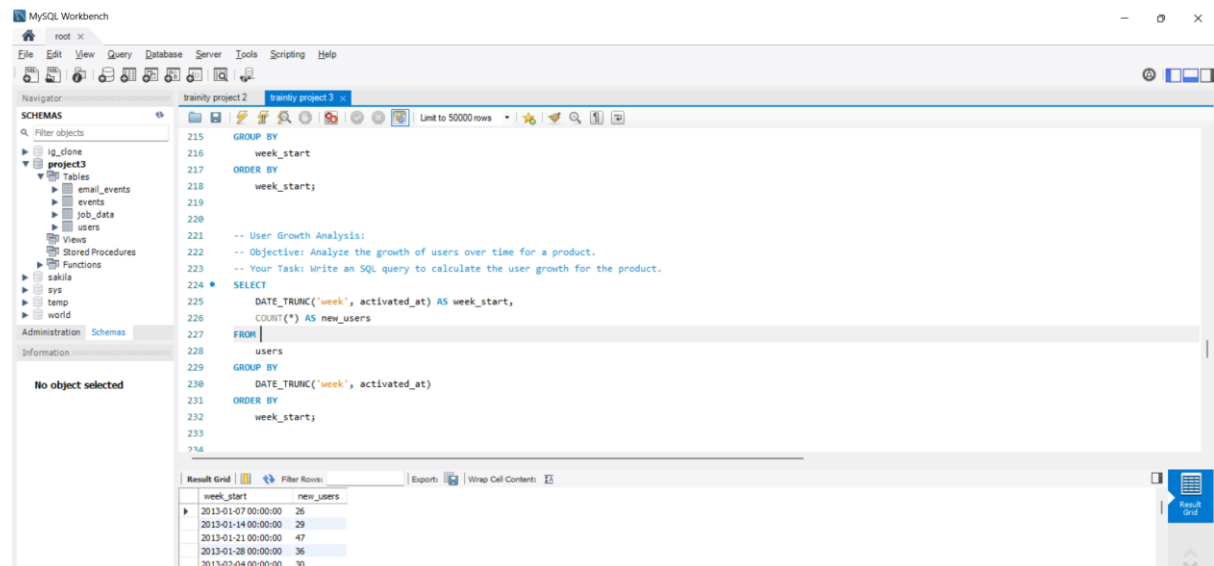
GROUP BY

```
DATE_TRUNC('week', activated_at)
```

ORDER BY

```
week_start;
```

Output:



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.

Code:

WITH cohort AS (

```
SELECT
```

```
    user_id,
```

```
    DATE_TRUNC('week', activated_at) AS cohort_week
```

```
FROM
```

```
    users
```

```
),
```

activity AS (

```
SELECT
```

```

        user_id,
        DATE_TRUNC('week', occurred_at) AS activity_week
    FROM
        events
),
cohort_activity AS (
    SELECT
        c.cohort_week,
        a.activity_week,
        COUNT(DISTINCT a.user_id) AS active_users
    FROM
        cohort c
    JOIN
        activity a ON c.user_id = a.user_id
    GROUP BY
        c.cohort_week, a.activity_week
)
SELECT
    cohort_week,
    activity_week,
    active_users,
    ROUND(100.0 * active_users / SUM(active_users) OVER (PARTITION BY cohort_week), 2) AS
retention_rate
FROM
    cohort_activity
ORDER BY
    cohort_week, activity_week;

```

Output:

The screenshot shows the MySQL Workbench interface. The left sidebar displays the 'SCHEMAS' tree with 'project3' selected. The main editor contains a SQL query for 'Weekly Retention Analysis'. The query uses a CTE named 'cohort' to calculate retention rates by joining 'users' and 'events' tables, grouping by cohort week and activity week. The bottom pane shows the 'Result Grid' with 5 rows of data.

```
233
234
235 -- Weekly Retention Analysis:
236 -- Objective: Analyze the retention of users on a weekly basis after signing up for a product.
237 -- Your Task: Write an SQL query to calculate the weekly retention of users based on their sign-up cohort.
238 select * from users;
239
240 WITH cohort AS (
241     SELECT
242         user_id,
243         DATE_TRUNC('week', activated_at) AS cohort_week
244     FROM
245         users
246 ),
247 activity AS (
248     SELECT
249         user_id,
250         DATE_TRUNC('week', occurred_at) AS activity_week
251     FROM
252         events
253 )
```

cohort_week	activity_week	active_users	retention_rate
2013-01-07 00:00:00	2014-05-05 00:00:00	2	4.26
2013-01-07 00:00:00	2014-05-12 00:00:00	3	6.38
2013-01-07 00:00:00	2014-05-19 00:00:00	3	6.38
2013-01-07 00:00:00	2014-05-26 00:00:00	3	6.38
2013-01-07 00:00:00	2014-06-02 00:00:00	2	4.26

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.

Code:

SELECT

week_start,

device,

COUNT(DISTINCT user_id) AS active_users,

COUNT(*) AS total_events,

AVG(event_count) AS avg_events_per_user

FROM (

SELECT

user_id,

device,

DATE_TRUNC('week', occurred_at) AS week_start,

COUNT(*) AS event_count

FROM

events

GROUP BY

```

        user_id, device, DATE_TRUNC('week', occurred_at)
    ) AS user_device_activity

GROUP BY

    week_start, device

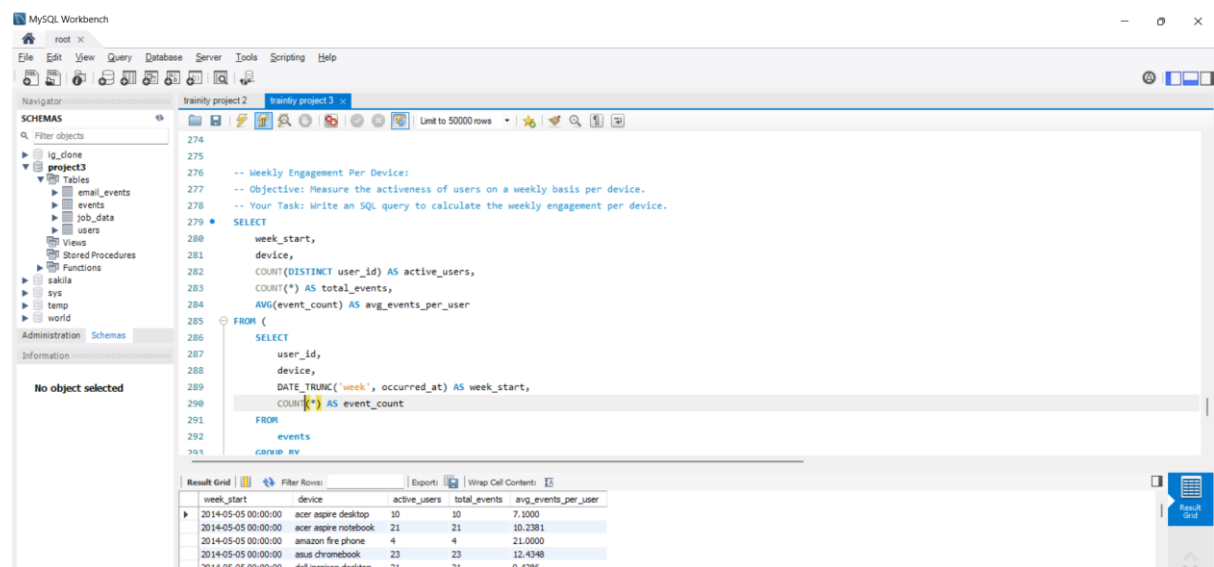
ORDER BY

    week_start, device

LIMIT 50000;

```

Output:



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.

Code:

```

SELECT

    DATE_TRUNC('week', occurred_at) AS week_start,

    action,

    COUNT(*) AS event_count,

    COUNT(DISTINCT user_id) AS unique_users

FROM

    email_events

```

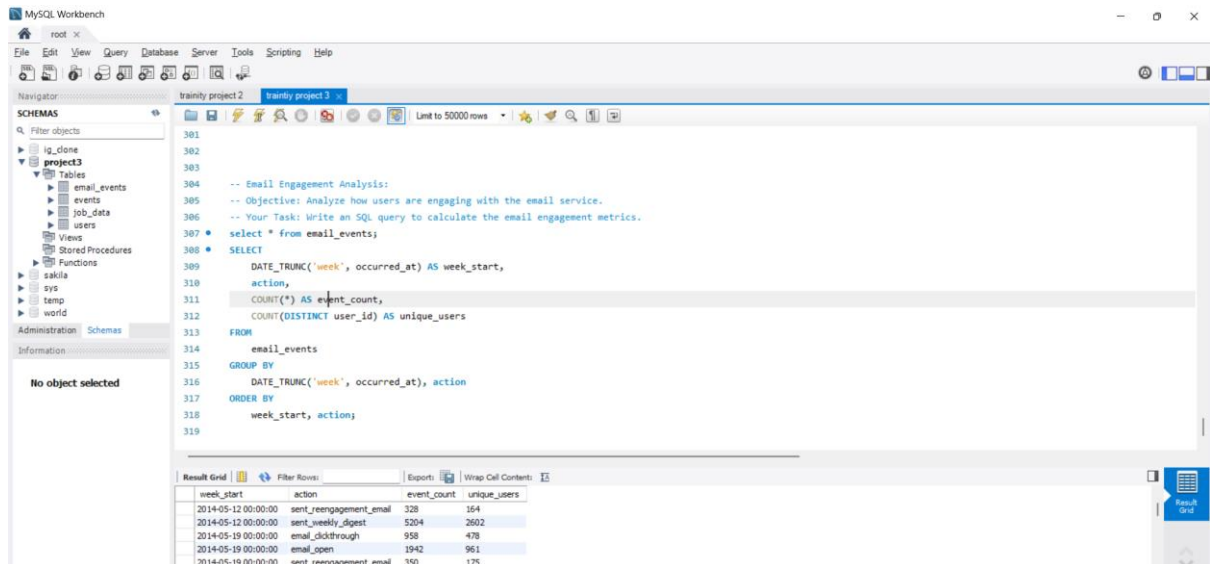
GROUP BY

DATE_TRUNC('week', occurred_at), action

ORDER BY

week_start, action;

Output:



The screenshot shows the MySQL Workbench interface. The left sidebar displays the 'SCHEMAS' panel with a tree view of databases including 'project3'. The main editor window contains a SQL query for email engagement analysis. The query uses DATE_TRUNC to group data by week and counts events and unique users. The results are shown in a table at the bottom of the window.

```
301
302
303
304 -- Email Engagement Analysis:
305 -- Objective: Analyze how users are engaging with the email service.
306 -- Your Task: Write an SQL query to calculate the email engagement metrics.
307 select * from email_events;
308
309 SELECT
310     DATE_TRUNC('week', occurred_at) AS week_start,
311     action,
312     COUNT(*) AS event_count,
313     COUNT(DISTINCT user_id) AS unique_users
314 FROM
315     email_events
316 GROUP BY
317     DATE_TRUNC('week', occurred_at), action
318 ORDER BY
319     week_start, action;
```

week_start	action	event_count	unique_users
2014-05-12 00:00:00	sent_reengagement_email	328	164
2014-05-12 00:00:00	sent_weekly_digest	5204	2602
2014-05-19 00:00:00	email_clickthrough	958	478
2014-05-19 00:00:00	email_open	1942	961
2014-05-19 00:00:00	sent_reengagement_email	350	175

Thank you