

Operation Analytics and Investigating Metric Spike

By Mukesh Dey



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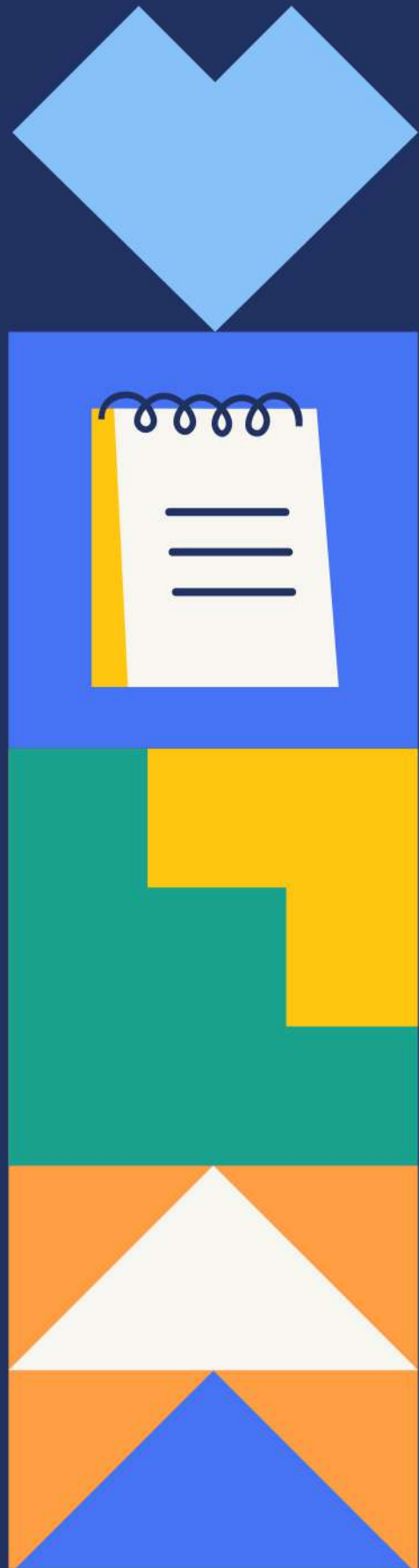
Job Data Analysis

- Jobs Reviewed Over Time
- Throughput Analysis
- Language Share Analysis
- Duplicate Rows Detection



Investigating Metric Spike

- Weekly User Engagement
- User Growth Analysis
- Weekly Retention Analysis
- Weekly Engagement Per Device
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Overview

- 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, I'll be working closely with various teams, such as operations, support, and marketing, helping them derive valuable insights from the data they collect.
- Here, my goal is to use advance SQL skills to analyze the data and provide valuable insights that can help improve the company's operations and understand sudden changes in key metrics.



Tech-Stack Used



Microsoft Excel



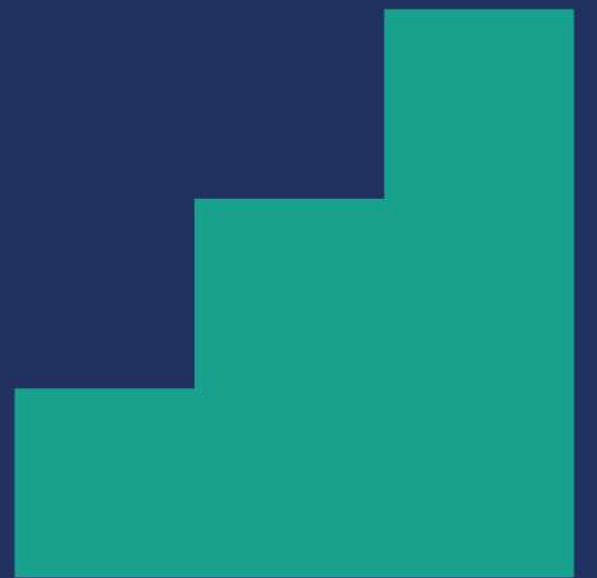
My SQL



Microsoft
Powerpoint

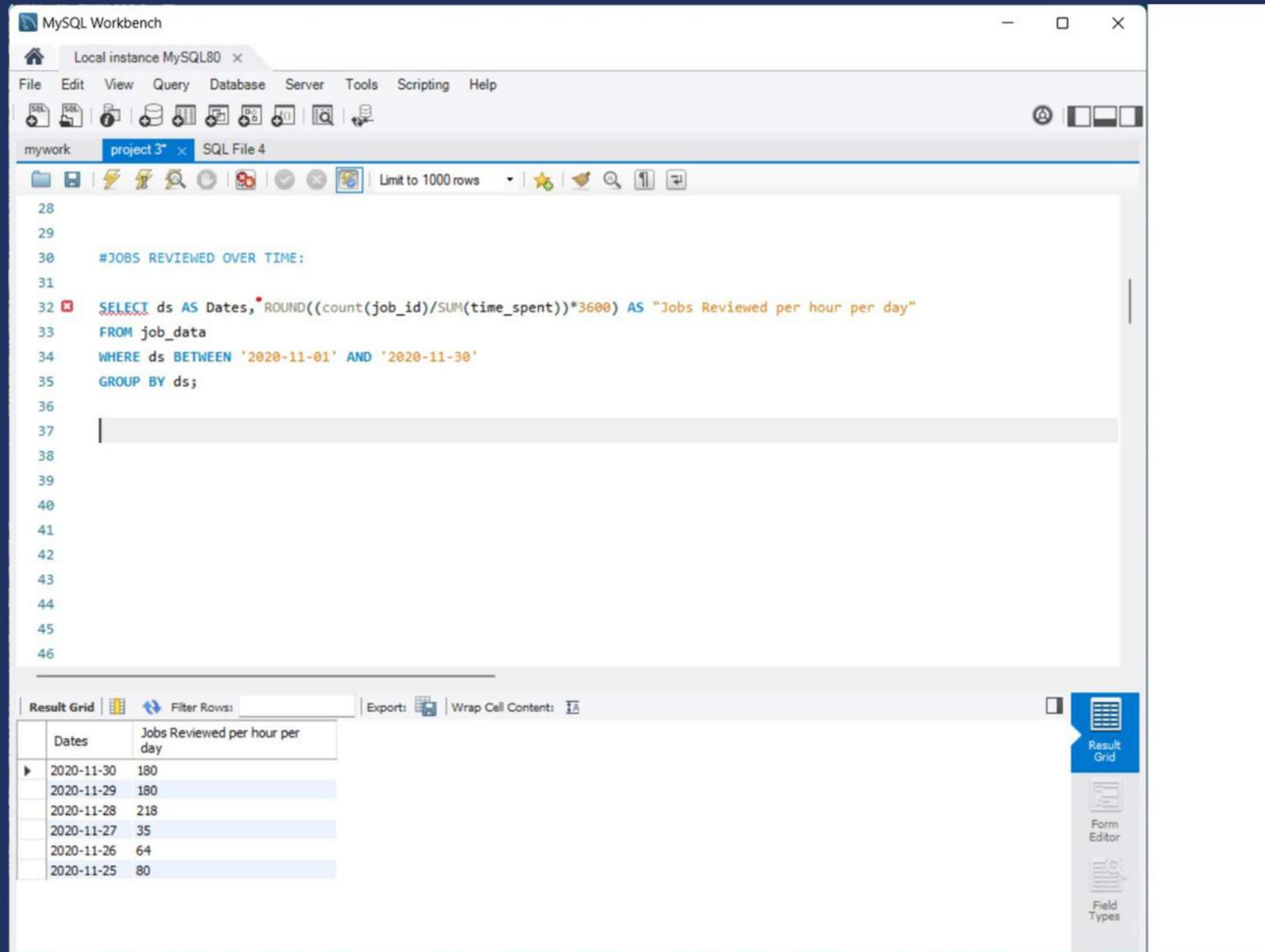


Job Data Analysis



- Jobs Reviewed Over Time

Here, my objective is to calculate the number of jobs reviewed per hour for each day in November 2020.

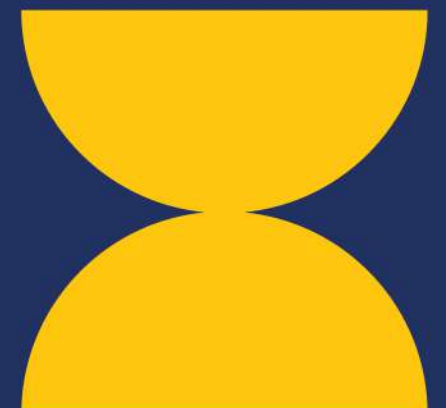


The screenshot shows the MySQL Workbench interface. The SQL editor contains the following query:

```
#JOBS REVIEWED OVER TIME:
SELECT ds AS Dates, ROUND((count(job_id)/SUM(time_spent))*3600) AS "Jobs Reviewed per hour per day"
FROM job_data
WHERE ds BETWEEN '2020-11-01' AND '2020-11-30'
GROUP BY ds;
```

The results are displayed in the Result Grid at the bottom:

Dates	Jobs Reviewed per hour per day
2020-11-30	180
2020-11-29	180
2020-11-28	218
2020-11-27	35
2020-11-26	64
2020-11-25	80



Insights

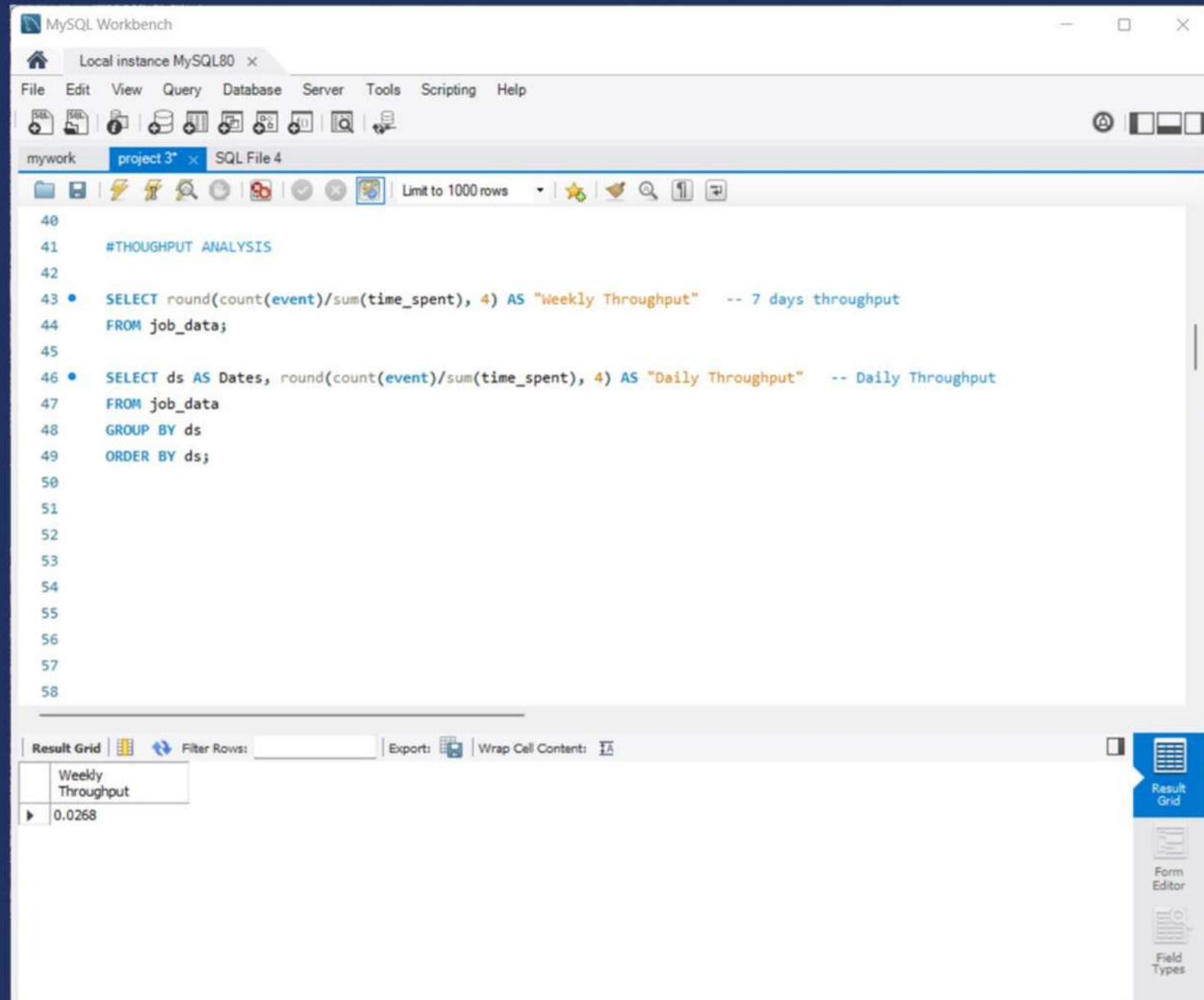
After analyzing the data it can be found that jobs reviewed per hour per day for the month of November 2020.

SQL Query used to find out the given output is mentioned in the previous slide.

Dates	Jobs Reviewed Per Hour
2020-11-30	180
2020-11-29	180
2020-11-28	218
2020-11-27	35
2020-11-26	64
2020-11-25	80

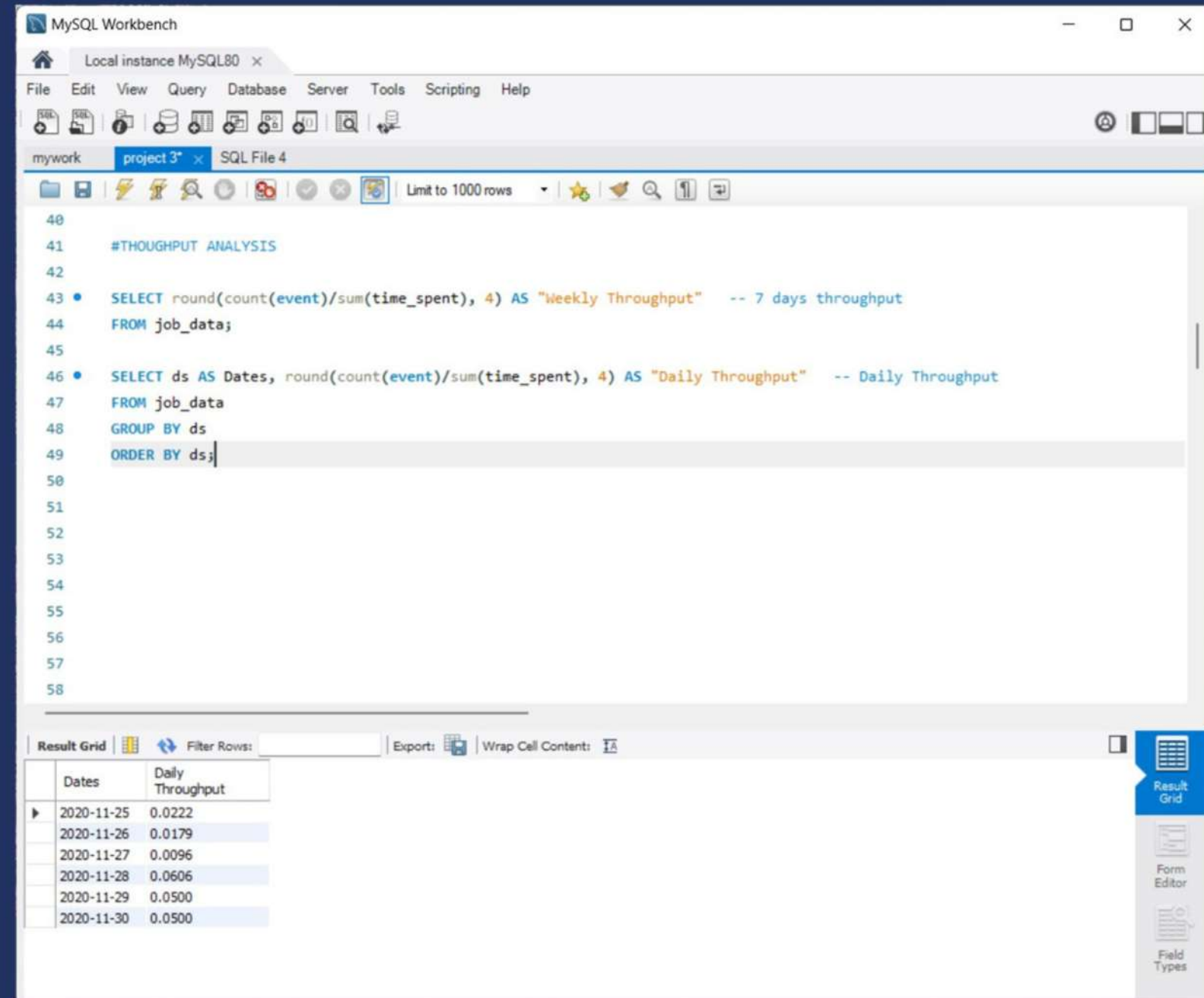
- Throughput Analysis

Here, my objective is to calculate the 7-day Rolling average of throughput analysis(number events per seconds



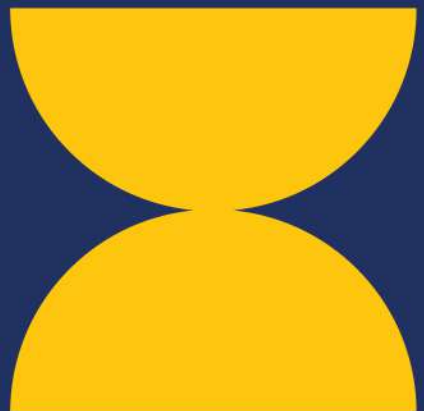
MySQL Workbench interface showing a SQL query for throughput analysis. The query is executed, and the result grid displays the following data:

Weekly Throughput
0.0268



MySQL Workbench interface showing a SQL query for throughput analysis. The query is executed, and the result grid displays the following data:

Dates	Daily Throughput
2020-11-25	0.0222
2020-11-26	0.0179
2020-11-27	0.0096
2020-11-28	0.0606
2020-11-29	0.0500
2020-11-30	0.0500



Insights

After analyzing the data it can be found that weekly throughput analysis is **0.0268**.

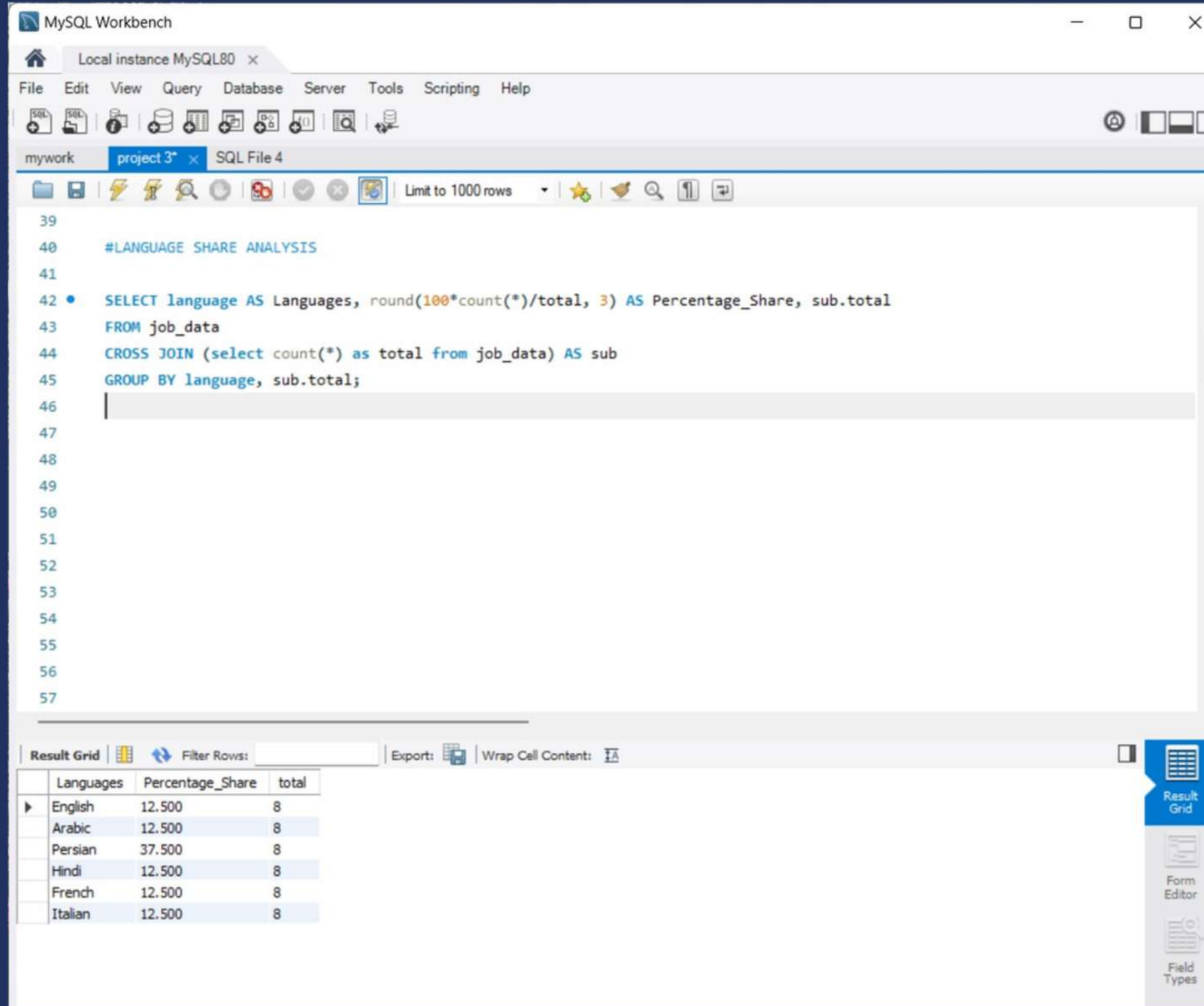
And if we see the daily throughput analysis the value comes as shown:

SQL Query used to find out the given output is mentioned in the previous slide.

Dates	Daily Throughout
2020-11-25	0.0222
2020-11-26	0.0179
2020-11-27	0.0096
2020-11-28	0.0606
2020-11-29	0.0500
2020-11-30	0.0500

- Language Share Analysis

Here, my objective is to calculate the percentage share of each language in the last 30 days.



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

```
39
40 #LANGUAGE SHARE ANALYSIS
41
42 • SELECT language AS Languages, round(100*count(*)/total, 3) AS Percentage_Share, sub.total
43 FROM job_data
44 CROSS JOIN (select count(*) as total from job_data) AS sub
45 GROUP BY language, sub.total;
46
47
48
49
50
51
52
53
54
55
56
57
```

The results are displayed in the Result Grid at the bottom:

Languages	Percentage_Share	total
English	12.500	8
Arabic	12.500	8
Persian	37.500	8
Hindi	12.500	8
French	12.500	8
Italian	12.500	8



Insights

The percentage share of each language over the last 30 days is given below:

Languages	Percentage_Share	Total
English	12.500	8
Arabic	12.500	8
Persian	37.500	8
Hindi	12.500	8
French	12.500	8
Italian	12.500	8

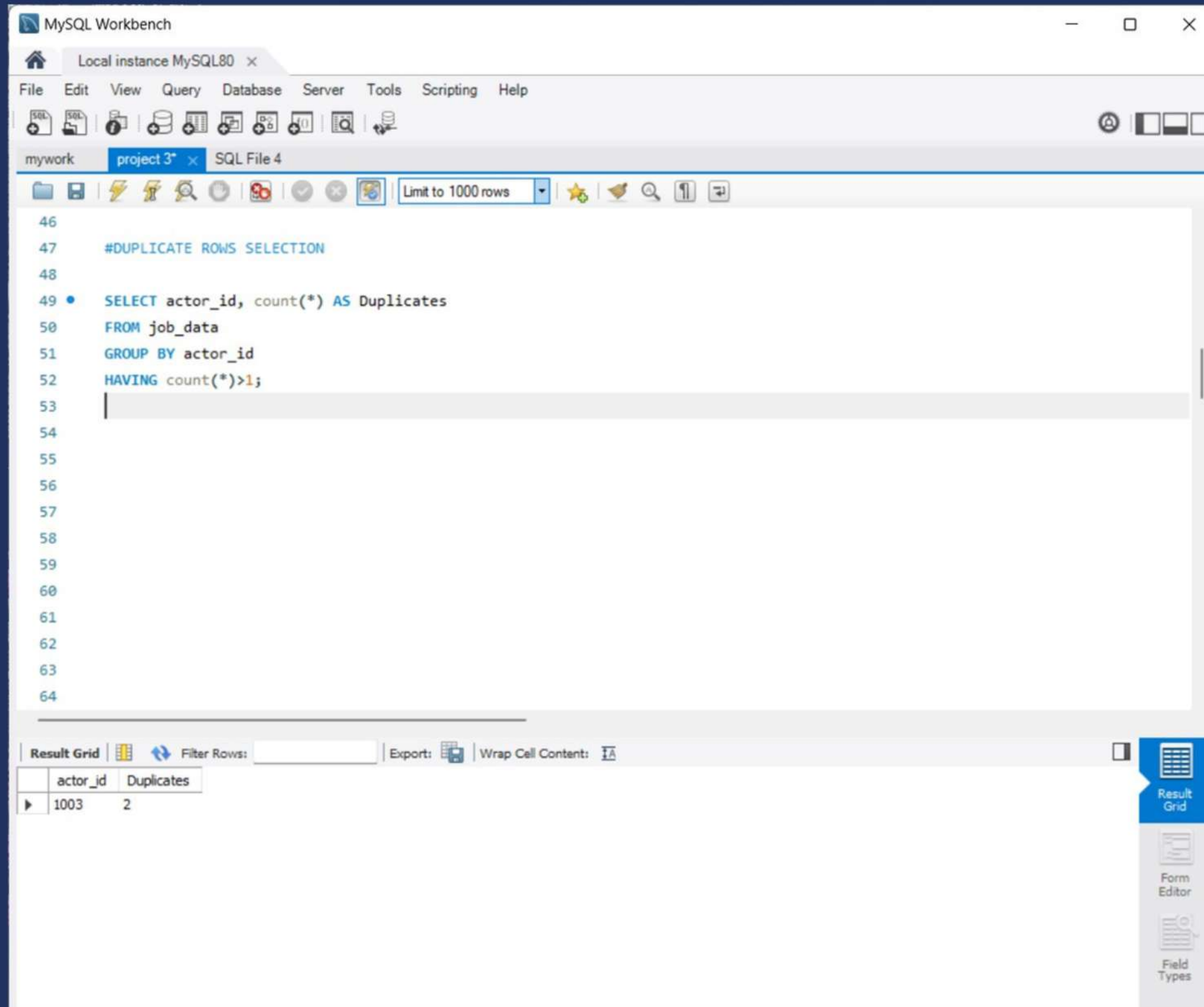


SQL Query used to find out the given output is mentioned in the previous slide.



- Duplicate Rows Detection

Here, my objective is to identify duplicate rows in the data.



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

```
46  
47 #DUPLICATE ROWS SELECTION  
48  
49 • SELECT actor_id, count(*) AS Duplicates  
50 FROM job_data  
51 GROUP BY actor_id  
52 HAVING count(*)>1;  
53  
54  
55  
56  
57  
58  
59  
60  
61  
62  
63  
64
```

The query is executed, and the Result Grid at the bottom shows the following data:

actor_id	Duplicates
1003	2



Insights

After analyzing the data it can be found that there are only **2 duplicate rows with the actor_id as "1003"**

SQL Query used to find out the given output is mentioned in the previous slide.



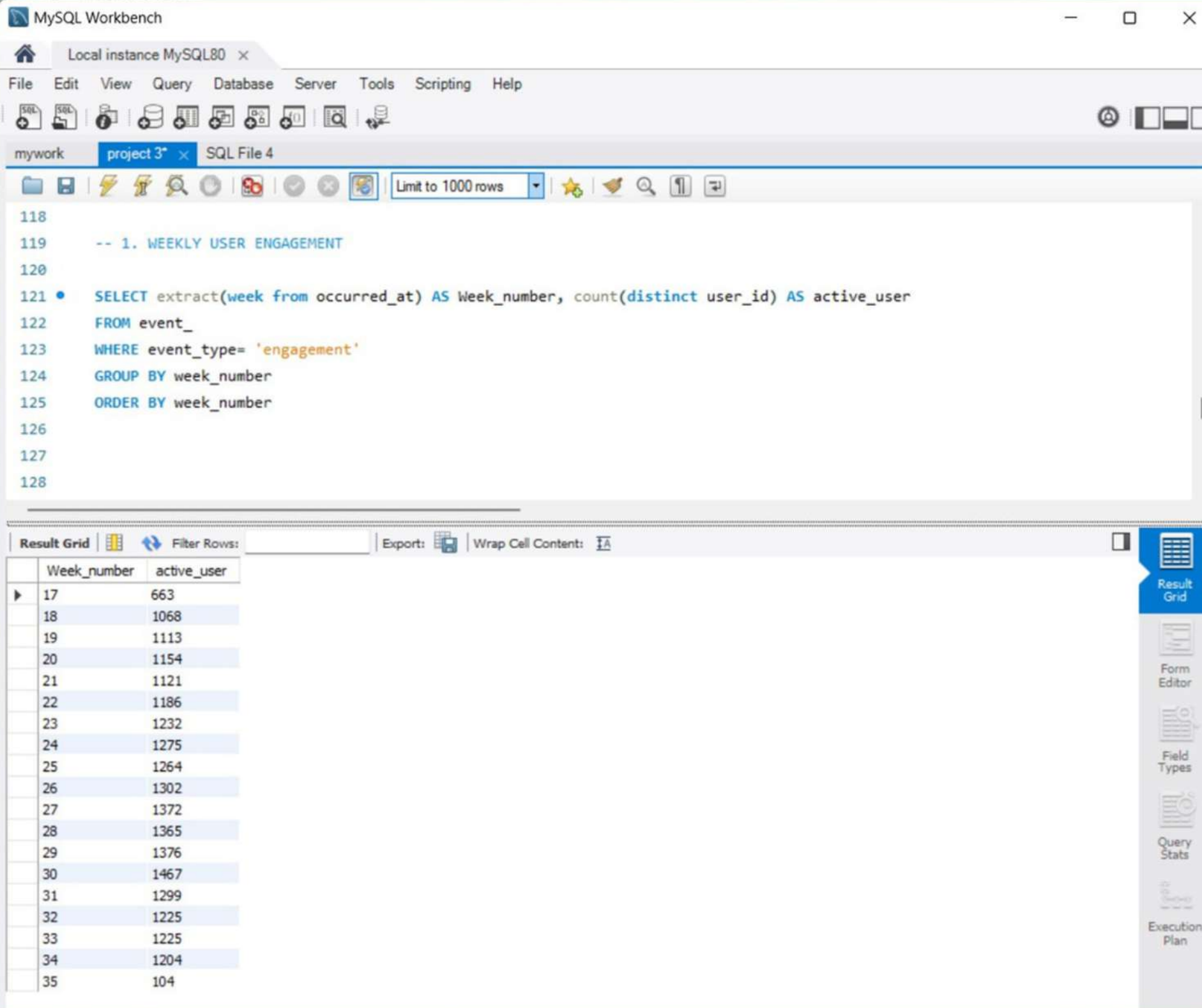


Investigation Metric Spikes



- Weekly User Engagement

Here, my objective is to measure the activeness of users on a weekly basis.



The screenshot shows the MySQL Workbench interface. The SQL editor contains a query to calculate weekly user engagement by counting distinct active users per week from an 'event_' table, filtered by 'engagement' event type. The results are displayed in a table with columns 'Week_number' and 'active_user'.

```
118
119 -- 1. WEEKLY USER ENGAGEMENT
120
121 • SELECT extract(week from occurred_at) AS Week_number, count(distinct user_id) AS active_user
122 FROM event_
123 WHERE event_type= 'engagement'
124 GROUP BY week_number
125 ORDER BY week_number
126
127
128
```

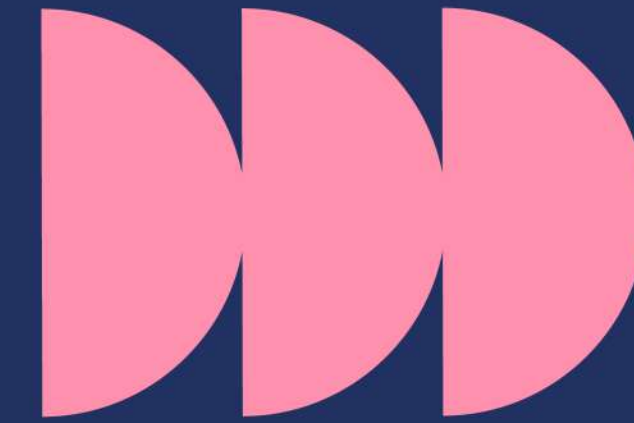
Week_number	active_user
17	663
18	1068
19	1113
20	1154
21	1121
22	1186
23	1232
24	1275
25	1264
26	1302
27	1372
28	1365
29	1376
30	1467
31	1299
32	1225
33	1225
34	1204
35	104



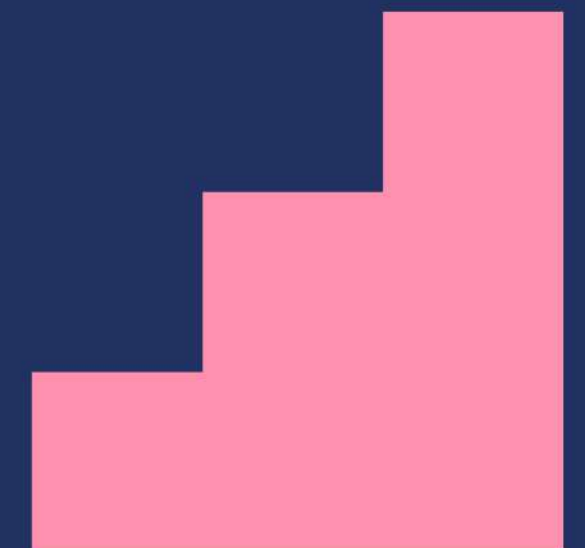
Insights

Week number and no. of active users according to that are mentioned below:

Week_number	Active_user
17	663
18	1068
19	1113
20	1154
21	1121
22	1186



SQL Query used to find out the given output is mentioned in the previous slide.



Insights

Week_number	Active_user
23	1232
24	1275
25	1264
26	1302
27	1372
28	1365
29	1376

Week_number	Active_user
30	1467
31	1299
32	1225
33	1225
34	1204
35	104

- User Growth Analysis:

Here, my objective is to analyze the growth of users over time for a product.

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

```
-- 2. USER GROWTH ANALYSIS

SELECT Months, Users, ROUND(((Users / LAG(Users, 1) OVER(ORDER BY Months) - 1) * 100), 4) AS 'Growth %'
FROM (SELECT EXTRACT(MONTH FROM created_at) AS Months, COUNT(created_at) AS Users
FROM users_
WHERE created_at IS NOT NULL
GROUP BY 1
ORDER BY 1) sub;
```

The results are displayed in the Result Grid below the query editor:

	Months	Users	Growth %
▶	1	712	NULL
	2	685	-3.7921
	3	765	11.6788
	4	907	18.5621
	5	993	9.4818
	6	1086	9.3656
	7	1281	17.9558
	8	1347	5.1522
	9	330	-75.5011
	10	390	18.1818
	11	399	2.3077
	12	486	21.8045



Insights

User growth is given according to the month from Jan-Dec according to the growth %. According, to that we can check in which month the growth was more or less.

SQL Query used to find out the given output is mentioned in the previous slide.



Months	Users	Growth %
1	712	NULL
2	685	-3.7921
3	765	11.6788
4	907	18.5621
5	993	9.4818
6	1086	9.3656
7	1281	17.9558
8	1347	5.1522

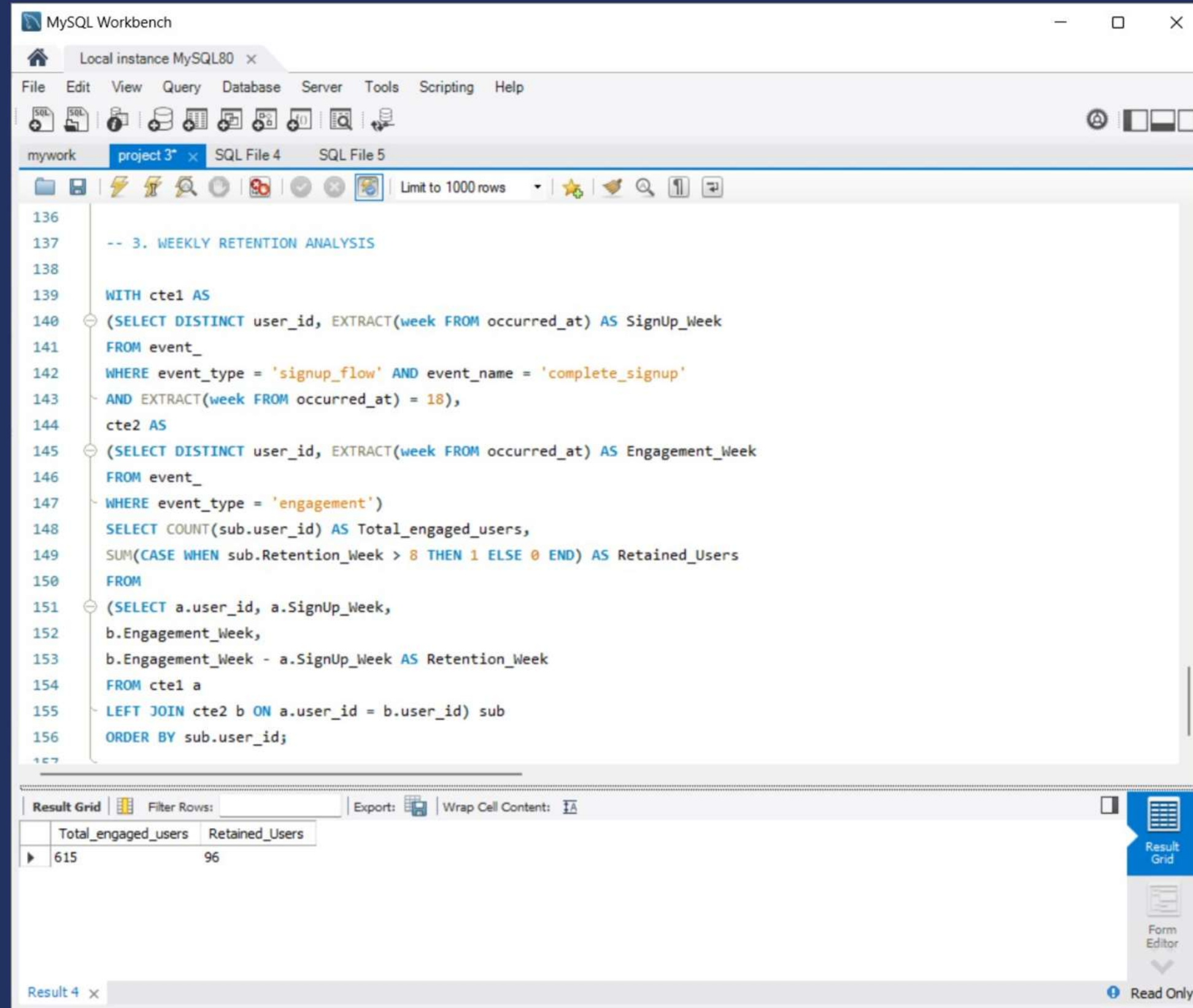
Insights



Months	Users	Growth %
9	330	-.75.5011
10	390	18.1818
11	399	2.3077
12	486	21.8045

- Weekly Retention Analysis:

Here, my objective is to measure the activeness of users on a weekly basis per device.



The screenshot displays the MySQL Workbench interface. The top menu bar includes File, Edit, View, Query, Database, Server, Tools, Scripting, and Help. Below the menu is a toolbar with various icons. The main window shows a SQL script for weekly retention analysis. The script defines two common table expressions (CTEs): 'cte1' for signups and 'cte2' for engagements. It then performs a left join between these two CTEs on user_id to calculate retention. The results are shown in a table at the bottom.

```
136
137  -- 3. WEEKLY RETENTION ANALYSIS
138
139  WITH cte1 AS
140  (SELECT DISTINCT user_id, EXTRACT(week FROM occurred_at) AS SignUp_Week
141   FROM event_
142   WHERE event_type = 'signup_flow' AND event_name = 'complete_signup'
143   AND EXTRACT(week FROM occurred_at) = 18),
144  cte2 AS
145  (SELECT DISTINCT user_id, EXTRACT(week FROM occurred_at) AS Engagement_Week
146   FROM event_
147   WHERE event_type = 'engagement')
148  SELECT COUNT(sub.user_id) AS Total_engaged_users,
149         SUM(CASE WHEN sub.Retention_Week > 8 THEN 1 ELSE 0 END) AS Retained_Users
150  FROM
151  (SELECT a.user_id, a.SignUp_Week,
152         b.Engagement_Week,
153         b.Engagement_Week - a.SignUp_Week AS Retention_Week
154   FROM cte1 a
155   LEFT JOIN cte2 b ON a.user_id = b.user_id) sub
156  ORDER BY sub.user_id;
```

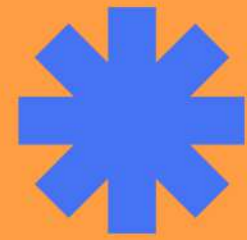
Total_engaged_users	Retained_Users
615	96



Insights

SQL Query which is used to analyze the data is given in the previous slide

The number of users calculated after analyzing the weekly data of users is given below:



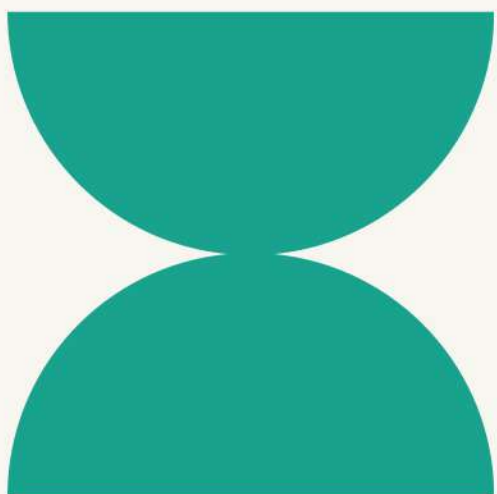
Total engaged users

The total no. of users which were engaged weekly after signing up for a product is
615



Retained users

The no. of users which were retained weekly based on the sign-up cohort is 96.



- Weekly Engagement Per Device:

Here, my objective is to measure the activeness of users on a weekly basis per device.

MySQL Workbench

Local instance MySQL80 x

File Edit View Query Database Server Tools Scripting Help

mywork* project 3* x SQL File 4 SQL File 5

Limit to 1000 rows

```
-- 4 WEEKLY ENGAGEMENT PER DEVICE
WITH cte AS
(SELECT EXTRACT(YEAR FROM occurred_at) || '-' || EXTRACT(WEEK FROM occurred_at) AS Week_number, device,
COUNT(DISTINCT user_id) AS user_count FROM event_ WHERE event_type = 'engagement' GROUP BY Week_number, device)
SELECT Week_number, device, user_count FROM cte
ORDER BY Week_number;
```

Result Grid | Filter Rows: | Export: | Wrap Cell Content: |

	Week_number	device	user_count
▶	1	acer aspire desktop	198
	1	acer aspire notebook	338
	1	amazon fire phone	89
	1	asus chromebook	355
	1	dell inspiron desktop	360
	1	dell inspiron notebook	677
	1	hp pavilion desktop	339
	1	htc one	196
	1	ipad air	478
	1	ipad mini	292
	1	iphone 4s	409
	1	iphone 5	1025
	1	iphone 5s	626
	1	kindle fire	205
	1	lenovo thinkpad	1309
	1	mac mini	150
	1	macbook air	950
	1	macbook pro	1952
	1	nexus 10	273
	1	nexus 5	621
	1	nexus 7	355
	1	nokia lumia 635	211
	1	samsung galaxy tablet	107
	1	samsung galaxy note	119
	1	samsung galaxy s4	803
	1	windows surface	182

Result 1 x



Insights

The activeness of users on weekly basis per device is found after analyzing the data from the table. And according to that the name of the device and no. of active users are mentioned below:

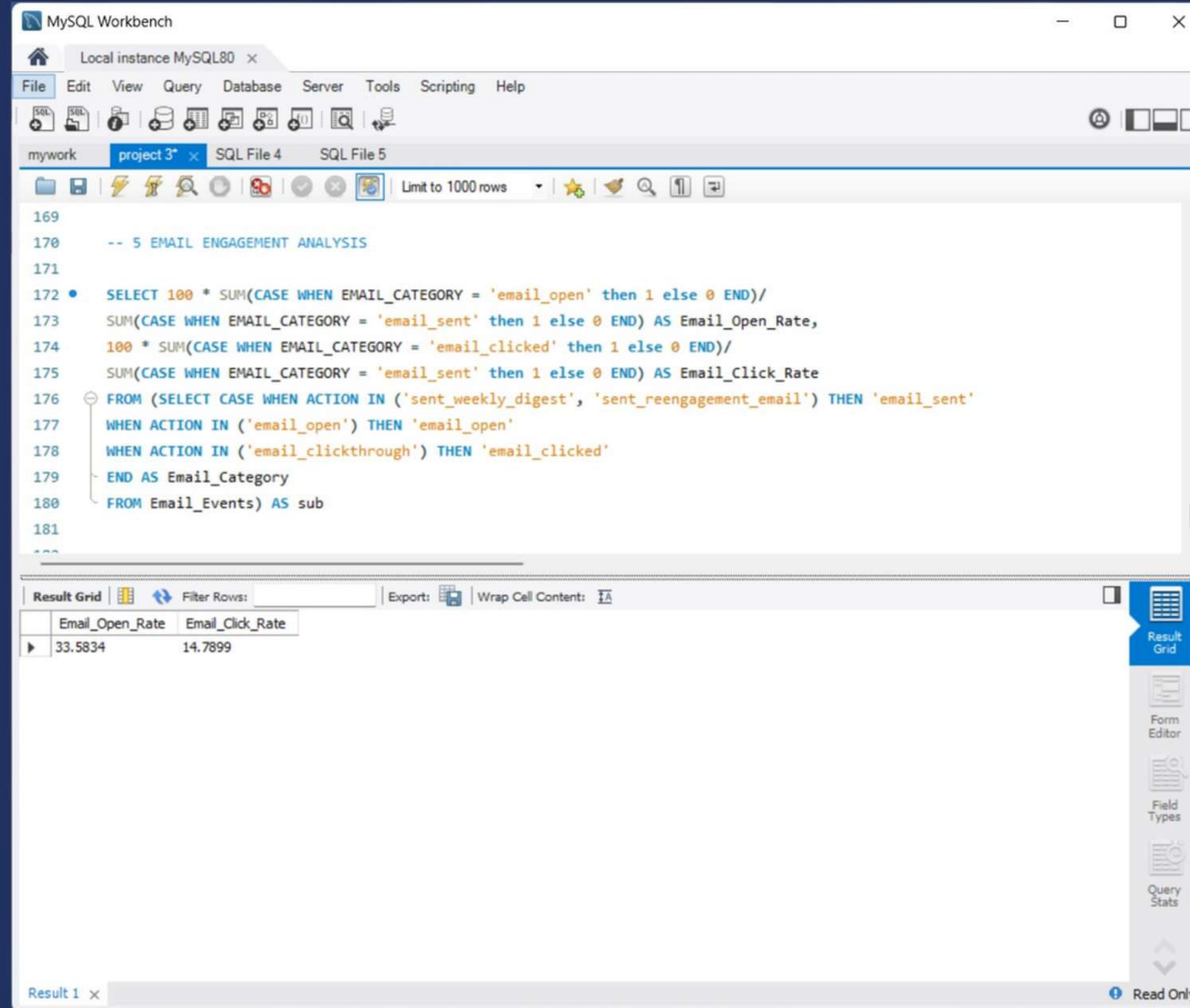
SQL query to analyze the find the result for the following objective is given in the previous slide.

Acer aspire notebook-338 ; Amazon fire phone- 89 ; Asus chromebook- 355 ; Dell inspiron desktop- 360; Dell inspiron notebook- 677 ; Hp pavilion desktop- 339 ; Htc pne- 196 ; Ipad air- 478 ; Ipad mini- 292 ; Iphone 4s- 409 ; Iphone 5- 1025 ; Kindle fire- 205 ; Lenovo thinkpad- 1309 ; Mac mini- 150 ; Macbook air- 950 ; Macbook pro- 1952 ; Nexus 10- 273 ; Nexus 5- 621 ; Nexus 7- 355 ; Nokia lumia 635- 211 ; Samsung galaxy tablet- 107 ; Samsung galaxy note- 119 ; Samsung galaxy s4- 803 ; Window surface- 182.



- Email Engagement Analysis:

Here, my objective is to analyze how users are engaged with the email service.



The screenshot shows the MySQL Workbench interface. The SQL editor contains a query to calculate email engagement rates. The query uses a subquery to categorize email actions into 'email_sent', 'email_open', and 'email_clicked'. The main query then calculates the 'Email_Open_Rate' and 'Email_Click_Rate' as percentages.

```
169
170 -- 5 EMAIL ENGAGEMENT ANALYSIS
171
172 • SELECT 100 * SUM(CASE WHEN EMAIL_CATEGORY = 'email_open' then 1 else 0 END)/
173     SUM(CASE WHEN EMAIL_CATEGORY = 'email_sent' then 1 else 0 END) AS Email_Open_Rate,
174     100 * SUM(CASE WHEN EMAIL_CATEGORY = 'email_clicked' then 1 else 0 END)/
175     SUM(CASE WHEN EMAIL_CATEGORY = 'email_sent' then 1 else 0 END) AS Email_Click_Rate
176 FROM (SELECT CASE WHEN ACTION IN ('sent_weekly_digest', 'sent_reengagement_email') THEN 'email_sent'
177     WHEN ACTION IN ('email_open') THEN 'email_open'
178     WHEN ACTION IN ('email_clickthrough') THEN 'email_clicked'
179     END AS Email_Category
180 FROM Email_Events) AS sub
181
182
```

The results are displayed in a table with two columns: Email_Open_Rate and Email_Click_Rate. The values are 33.5834 and 14.7899 respectively.

Email_Open_Rate	Email_Click_Rate
33.5834	14.7899

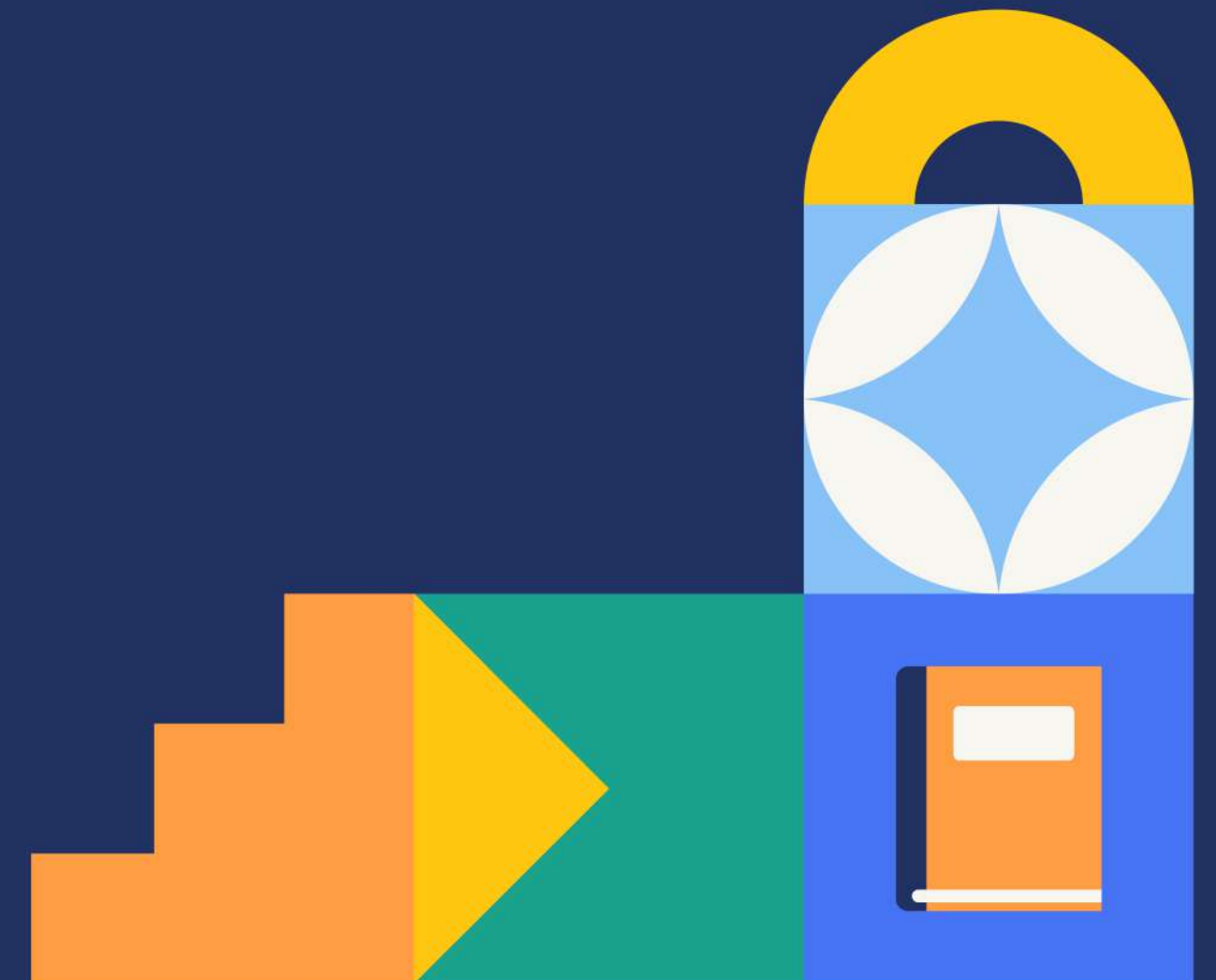


Insights

After analyzing the data from the table Email_events table the activeness of users how they were engaged with email service has been found as follows:

- Email Open Rate :- 33.5834
- Email Click Rate :- 14.7899

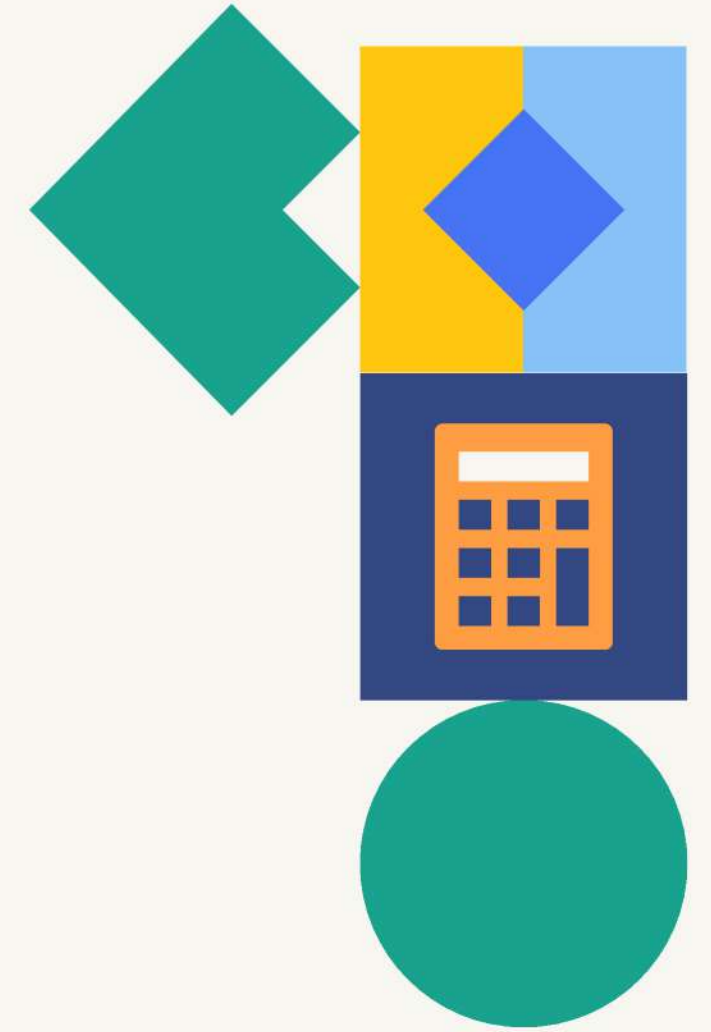
SQL query used to analyze the following objective is mentioned in the previous slide



Result

After analyzing the data from the information given into tables, various analysis has been made based upon which it will be easy to find the data which company wants for their employees as well as their product.

And by analyzing the data I'm able to learn advanced SQL and importing data directly from MS Excel to My SQL.



Thank you 

