

Operation Analytics and Investigating Metric Spike

A Project Delivered By Using
SQL techniques.



Presented By : Jacky Kumar



PROJECT DESCRIPTION:

Operation analytics is done for the complete end-to-end operations of the company. With its help, the company then finds areas it needs to improve. You work closely with the ops, support, and marketing team among others, and help them derive insight from the data they collect.

Being one of the most important analysis of a company, this is further used to predict overall growth or decline in fortune. This means better automation, better understanding between cross-functional teams, and more effective workflows.



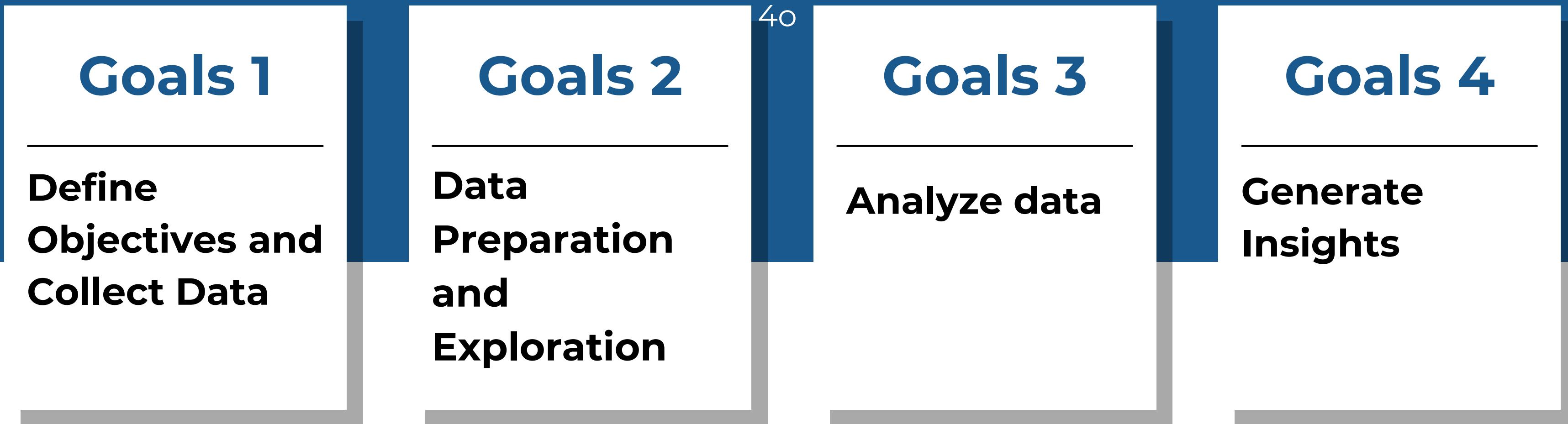
Investigation of metric spike

This is an important part of operation analytics because, as a Data Analyst, you must be able to understand-or make other teams understand-questions such as: Why is daily engagement dipping? Why have sales taken a dip? Etc. Such questions come up daily and, for that, investigating metric spikes becomes very important.



APPROACH

To approach an Operation & Metric Analytics project, start by defining clear objectives and identifying the key metrics (KPIs) that align with business goals. Collect relevant operational data from various sources to ensure a comprehensive dataset for analysis.



TECH-STACK USED

This software provides better data security, more Built-in functions, and on-demand scalability; apart from this, it also provides an auto-completing feature that makes it easy for the developer to write the queries.

- MySQL Workbench 8.0

- Microsoft Excel for importing and exporting the data



MY INSIGHTS

- While making this project, I came to learn about SQL, how to implement the queries, and various in-built functions that can be used to get the data.
- I got a good exposure to MySQL, respectively, on how SQL and the functions could be used in analysing the data from the dataset, how the queries work and are executed, and how I am playing a major role in retrieving the data instantly without searching for it in the dataset.

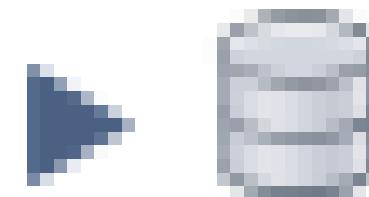


DATABASE CREATION

Code

```
create database project3;  
use project3;
```

Output



project3

creating a table and importing data in the table

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', 'Aeabic', 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');
```

Output

	ds	job_id	actor_id	event	language	time_spent	org
▶	2020-11-30	21	1001	skip	English	15	A
	2020-11-30	22	1006	transfer	Aeabic	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

Case Study 01.

Task 01. Number of jobs reviewed: Amount of job review over time.

Code

```
select ds as date,  
count(job_id) as joint_job_id,  
round((sum(time_spent) / 3600), 2) as total_time_per_hour,  
round((count(job_id) / (sum(time_spent)/3600)),2) as job_review  
from job_data  
where  
ds between '2020-11-01' and '2020-11-30'  
group by ds  
order by ds;
```

Output

	date	joint_job_id	total_time_per_hour	job_review
▶	2020-11-25	1	0.01	80.00
	2020-11-26	1	0.02	64.29
	2020-11-27	1	0.03	34.62
	2020-11-28	2	0.01	218.18
	2020-11-29	1	0.01	180.00
	2020-11-30	2	0.01	180.00

Insights

We can say that 0.01 jobs reviewed per hour in nov 2020.

On 28th nov 2020 was highest job reviewed which was 218.18

Task 02. Throughput Analysis

Code

```
select round(count(event) / sum(time_spent), 2) As weekly_avg_throughput  
from job_data;
```

```
select ds as dates,  
round(count(event) / sum(time_spent),2) as daily_avg_throughput  
from job_data  
group by ds  
order by ds;
```

Output

	weekly_avg_throughput	
▶	0.03	
	dates	daily_avg_throughput
▶	2020-11-25	0.02
	2020-11-26	0.02
	2020-11-27	0.01
	2020-11-28	0.06
	2020-11-29	0.05
	2020-11-30	0.05

explain whether you prefer using the daily metric or the 7-day rolling average for throughput, and why.

- 7-Day Rolling Average: (Use Case: Ideal for long-term planning and performance evaluation.)
- Daily Metric: (Use Case: Suitable for monitoring real-time or daily performance and identifying anomalies.)

Conclusion: I would prefer the 7-day rolling average for throughput analysis because it reduces noise from daily variations and gives a clearer picture of overall trends. However, for addressing specific anomalies or urgent issues, daily metrics are more actionable.

Task 03. Language Share Analysis:

Code

```
select language, round(100* count(*) / total, 2) as percentage,  
jd.total from job_data  
cross join (select count(*) as total  
from job_data) as jd  
group by language, jd.total;
```

Output

	language	percentage	total
▶	English	12.50	8
	Aeabic	12.50	8
	Persian	37.50	8
	Hindi	12.50	8
	French	12.50	8
	Italian	12.50	8

Insights

We can say that Persian language is the highest language used with the number of percentage share of 37.50 incompared to other languages.

Task 04. Duplicate Rows Detection:

Code

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

	actor_id	duplicate
▶	1003	2

Insights

Total of 8 rows, we have only 2 duplicate rowsThe

actor_id 1003 is the duplicate.

CASE STUDY 02.

Output

Task 01. Weekly User Engagement:

Code

```
select extract(week from occurred_at) as week_num,  
       count(distinct user_id) as active_users  
  from events  
 where event_type = 'engagement'  
 group by week_num  
 order by week_num;
```

Insights

01. The highest engagement in 30th weeks, with 1467 users.
02. The minimum engagement in 35th weeks, with 104 users.

	week_num	active_users
▶	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

Output

Task 02. User Growth Analysis:

Code

```
select count(*) from users where state = 'active';
select distinct state from users;

with weekly_active_users as(
    select extract(year from created_at) as year,
    extract(week from created_at) as week_number,
    count(distinct user_id) num_of_users
    from users
    group by year, week_number
)

select year, week_number, num_of_users,
sum(num_of_users) over (order by year, week_number) as cumulative_users
from weekly_active_users
order by year, week_number;
```

year	week_num	num_of_u	cumulative	year	26	56	1131	2014	0	83	3366	2014	26	201	7509
2013	0	23	23	2013	27	52	1183	2014	1	126	3492	2014	27	222	7731
2013	1	30	53	2013	28	72	1255	2014	2	109	3601	2014	28	215	7946
2013	2	48	101	2013	29	67	1322	2014	3	113	3714	2014	29	221	8167
2013	3	36	137	2013	30	67	1389	2014	4	130	3844	2014	30	238	8405
2013	4	30	167	2013	31	67	1456	2014	5	133	3977	2014	31	193	8598
2013	5	48	215	2013	32	71	1527	2014	6	135	4112	2014	32	245	8843
2013	6	38	253	2013	33	73	1600	2014	7	125	4237	2014	33	261	9104
2013	7	42	295	2013	34	78	1678	2014	8	129	4366	2014	34	259	9363
2013	8	34	329	2013	35	63	1741	2014	9	133	4499	2014	35	18	9381
2013	9	43	372	2013	36	72	1813	2014	10	154	4653	2014	36	207	7308
2013	10	32	404	2013	37	85	1898	2014	11	130	4783	2014	37	163	5936
2013	11	31	435	2013	38	90	1988	2014	12	148	4931	2014	38	185	6121
2013	12	33	468	2013	39	84	2072	2014	13	167	5098	2014	39	202	6297
2013	13	39	507	2013	40	87	2159	2014	14	162	5260	2014	40	220	6480
2013	14	35	542	2013	41	73	2232	2014	15	164	5424	2014	41	238	6676
2013	15	43	585	2013	42	99	2331	2014	16	179	5603	2014	42	256	6872
2013	16	46	631	2013	43	89	2420	2014	17	170	5773	2014	43	274	7101
2013	17	49	680	2013	44	96	2516	2014	18	163	5936	2014	44	292	7308
2013	18	44	724	2013	45	91	2607	2014	19	185	6121	2014	45	310	7509
2013	19	57	781	2013	46	88	2695	2014	20	176	6297	2014	46	328	7731
2013	20	39	820	2013	47	102	2797	2014	21	183	6480	2014	47	346	7946
2013	21	49	869	2013	48	97	2894	2014	22	196	6676	2014	48	364	8167
2013	22	54	923	2013	49	116	3010	2014	23	196	6872	2014	49	382	8405
2013	23	50	973	2013	50	124	3134	2014	24	229	7101	2014	50	400	8598
2013	24	45	1018	2013	51	102	3236	2014	25	207	7308	2014	51	418	8843
2013	25	57	1075	2013	52	47	3283	2014				2014			

Insights

Highest number of users was in the 33rd weeks of 2014 with 261 new users reaching a total of 9104 users.

Task 03. Weekly Retention Analysis: Code

SELECT

```
first AS "week_numbers",
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
    m.user_id,
    m.login_week,
    n.first,
    m.login_week - n.first AS week_number
FROM (
```

SELECT

```
    user_id,
    EXTRACT(WEEK FROM occurred_at) AS login_week
FROM
```

events

GROUP BY

user_id,
login_week

) m

) JOIN (

SELECT

user_id,
 MIN(EXTRACT(WEEK FROM occurred_at)) AS first

FROM

events

GROUP BY

user_id

) n

ON m.user_id = n.user_id

) sub

GROUP BY

first

ORDER BY

first;

Output

week	week_17	week_18	week_19	week_20	week_21	week_22	week_23	week_24	week_25	week_26	week_27	week_28	week_29	week_30	week_31	week_32	week_33	week_34	week_35	week_18
17	663	472	324	251	205	187	167	146	145	145	136	131	132	143	116	91	82	77	5	
18	596	362	261	203	168	147	144	127	113	122	106	118	127	110	97	85	67	4	0	
19	427	284	173	153	114	95	91	81	95	82	68	65	63	42	51	49	2	0	0	
20	358	223	165	121	91	72	63	67	63	65	67	41	40	33	40	0	0	0	0	
21	317	187	131	91	74	63	75	72	58	48	45	39	35	28	2	0	0	0	0	
22	326	224	150	107	87	73	63	60	55	48	41	39	31	1	0	0	0	0	0	
23	328	219	138	101	90	79	69	61	54	47	35	30	0	0	0	0	0	0	0	
24	339	205	143	102	81	63	65	61	38	39	29	0	0	0	0	0	0	0	0	
25	305	218	139	101	75	63	50	46	38	35	2	0	0	0	0	0	0	0	0	
26	288	181	114	83	73	55	47	43	29	0	0	0	0	0	0	0	0	0	0	
27	292	199	121	106	68	53	40	36	1	0	0	0	0	0	0	0	0	0	0	
28	274	194	114	69	46	30	28	3	0	0	0	0	0	0	0	0	0	0	0	
29	270	186	102	65	47	40	1	0	0	0	0	0	0	0	0	0	0	0	0	
30	294	202	121	78	53	3	0	0	0	0	0	0	0	0	0	0	0	0	0	
31	215	145	76	57	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
32	267	188	94	8	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
33	286	202	9	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
34	279	44	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
35	18	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	

Insights

1. The users joined in the 17th week shows the longest period of users retention in 18th week.
2. The lowest users joined in 35th week is 18 users.

Task 04. Weekly Retention Analysis:

Code

SELECT

```
EXTRACT(WEEK FROM occurred_at) AS week_number,  
COUNT(DISTINCT CASE WHEN device = 'dell inspiron notebook' THEN user_id ELSE NULL END) AS dell_inspiron_noteboo  
COUNT(DISTINCT CASE WHEN device = 'iphone 5' THEN user_id ELSE NULL END) AS iphone_5,  
COUNT(DISTINCT CASE WHEN device = 'iphone 4s' THEN user_id ELSE NULL END) AS iphone_4s,  
COUNT(DISTINCT CASE WHEN device = 'iphone 5s' THEN user_id ELSE NULL END) AS iphone_5s,  
COUNT(DISTINCT CASE WHEN device = 'ipad air' THEN user_id ELSE NULL END) AS ipad_air,  
COUNT(DISTINCT CASE WHEN device = 'windows surface' THEN user_id ELSE NULL END) AS windows_surface,  
COUNT(DISTINCT CASE WHEN device = 'macbook air' THEN user_id ELSE NULL END) AS macbook_air,  
COUNT(DISTINCT CASE WHEN device = 'macbook pro' THEN user_id ELSE NULL END) AS macbook_pro,  
COUNT(DISTINCT CASE WHEN device = 'ipad mini' THEN user_id ELSE NULL END) AS ipad_mini,  
COUNT(DISTINCT CASE WHEN device = 'kindle fire' THEN user_id ELSE NULL END) AS kindle_fire,  
COUNT(DISTINCT CASE WHEN device = 'amazon fire phone' THEN user_id ELSE NULL END) AS amazon_fire_phone,  
COUNT(DISTINCT CASE WHEN device = 'nexus 5' THEN user_id ELSE NULL END) AS nexus_5,  
COUNT(DISTINCT CASE WHEN device = 'nexus 7' THEN user_id ELSE NULL END) AS nexus_7,  
COUNT(DISTINCT CASE WHEN device = 'nexus 10' THEN user_id ELSE NULL END) AS nexus_10,  
COUNT(DISTINCT CASE WHEN device = 'samsung_galaxy_s4' THEN user_id ELSE NULL END) AS samsung_galaxy_s4,  
COUNT(DISTINCT CASE WHEN device = 'samsung_galaxy_tablet' THEN user_id ELSE NULL END) AS samsung_galaxy_tablet,  
COUNT(DISTINCT CASE WHEN device = 'samsung_galaxy_note' THEN user_id ELSE NULL END) AS samsung_galaxy_note,  
COUNT(DISTINCT CASE WHEN device = 'lenovo thinkpad' THEN user_id ELSE NULL END) AS lenovo_thinkpad,  
COUNT(DISTINCT CASE WHEN device = 'acer aspire notebook' THEN user_id ELSE NULL END) AS acer_aspire_notebook,  
COUNT(DISTINCT CASE WHEN device = 'asus chromebook' THEN user_id ELSE NULL END) AS asus_chromebook,  
COUNT(DISTINCT CASE WHEN device = 'htc one' THEN user_id ELSE NULL END) AS htc_one,  
COUNT(DISTINCT CASE WHEN device = 'nokia lumia 635' THEN user_id ELSE NULL END) AS nokia_lumia_635,  
COUNT(DISTINCT CASE WHEN device = 'mac mini' THEN user_id ELSE NULL END) AS mac_mini,  
COUNT(DISTINCT CASE WHEN device = 'hp pavilion desktop' THEN user_id ELSE NULL END) AS hp_pavilion_desktop,  
COUNT(DISTINCT CASE WHEN device = 'dell inspiron desktop' THEN user_id ELSE NULL END) AS dell_inspiron_desktop
```

FROM

```
events
```

WHERE

WHERE

```
event_type = 'engagement'
```

GROUP BY

```
week_number
```

ORDER BY

```
week_number;
```

Output

week	dell	iphc	iph	iph	ipa	wind	mac	mac	ipa	kin	am	nex	nex	nex	sai	sai	sai	leno	ace	asi	htc	nok	ma	hp	dell
17	46	65	21	42	27	10	54	143	19	6	4	40	18	16	0	0	0	86	20	21	16	17	6	14	18
18	77	113	46	73	52	10	121	252	30	27	9	73	30	30	0	0	0	153	33	42	19	33	13	37	58
19	83	115	44	79	55	16	112	266	36	21	12	87	41	25	0	0	0	178	41	27	30	23	18	40	36
20	84	125	55	79	59	21	119	256	32	23	11	103	32	22	0	0	0	173	40	41	29	22	26	30	52
21	80	137	45	74	51	17	110	247	23	30	5	91	29	25	0	0	0	167	47	38	21	25	18	44	41
22	92	125	45	71	58	15	145	251	34	21	5	96	45	27	0	0	0	176	41	52	24	25	25	38	52
23	103	152	53	79	41	14	124	266	33	25	16	88	36	45	0	0	0	176	43	49	20	31	18	54	53
24	99	142	53	79	57	22	152	255	39	25	11	87	49	38	0	0	0	165	40	43	20	35	29	56	59
25	105	137	40	78	57	22	121	275	30	24	13	89	51	29	0	0	0	197	47	38	21	37	21	52	52
26	89	152	50	94	56	21	134	269	43	26	13	87	46	29	0	0	0	192	35	49	23	42	11	46	60
27	89	163	67	83	55	33	142	302	35	25	10	84	40	37	0	0	0	202	49	52	27	31	15	56	53
28	103	151	61	93	54	33	148	295	35	31	6	85	39	26	0	0	0	220	49	50	26	35	28	56	56
29	113	144	60	90	52	28	148	295	34	37	12	77	45	25	0	0	0	209	53	49	31	43	31	58	54
30	127	152	65	##	70	19	159	322	35	25	12	84	62	36	0	0	0	206	60	56	31	34	23	42	54
31	113	135	56	71	55	19	147	321	27	14	14	69	38	24	0	0	0	207	55	56	13	28	24	51	44
32	104	119	34	67	48	10	125	307	30	12	12	67	25	30	0	0	0	179	55	62	18	28	20	51	57
33	110	110	35	65	40	15	133	312	28	14	14	70	30	23	0	0	0	191	46	49	19	27	32	38	37
34	105	101	50	70	39	18	136	292	25	13	11	70	33	25	0	0	0	193	63	47	25	17	30	36	49
35	9	2	6	3	0	3	10	17	2	3	0	4	2	2	0	0	0	16	3	6	2	2	2	1	1

Insights

01. The most users can use macbook pro(322 users on 30th weeks, and lenovo thinkpad(220 users on 28th weeks) also iphone_5 (163 users on the 27th weeks).

Task 05. Email Engagement Analysis:

Code

```
SELECT  
    100.0 * SUM(CASE WHEN email_action = 'email-open' THEN 1 ELSE 0 END) /  
    NULLIF(SUM(CASE WHEN email_action = 'email-sent' THEN 1 ELSE 0 END), 0) AS email_open_rate,  
  
    100.0 * SUM(CASE WHEN email_action = 'email-clicked' THEN 1 ELSE 0 END) /  
    NULLIF(SUM(CASE WHEN email_action = 'email-sent' THEN 1 ELSE 0 END), 0) AS email_clicked_rate  
FROM (  
    SELECT *  
    ,  
    CASE  
        WHEN action IN ('sent_weekly_digest', 'sent_reengagement_email') THEN 'email-sent'  
        WHEN action = 'email_open' THEN 'email-open'  
        WHEN action = 'email_clickthrough' THEN 'email-clicked'  
        ELSE NULL  
    END AS email_action  
    FROM  
        project3.email_events  
) aj
```

Output

	email_open_rate	email_clicked_rate
▶	33.58339	14.78989

Insights

01. All emails sent around 33% were opened and 14% were only clicked..



THANK YOU