

# PROJECT - 3

## Operation Analytics and Investigating Metric Spike

Advanced SQL

### Case Study 1: Job Data Analysis

#### Task:

##### 1. Jobs Reviewed Over Time:

Objective: Calculate the number of jobs reviewed per hour for each day in November 2020.

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

**Query:** select avg(t) as 'avg jobs reviewed per day per hour',  
avg(p) as 'avg jobs reviewed per day per second'  
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) a;

#### Output:

avg jobs reviewed per day per hour

**126.18048333**

avg jobs reviewed per day  
Per second

**0.03505000**

## 2. Throughput Analysis:

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

**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.

**Query:** `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:**

Dates	Daily 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

### 3. Language Share Analysis:

Objective: Calculate the percentage share of each language in the last 30 days.

**Task:** Write an SQL query to calculate the percentage share of each language over the last 30 days.

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**Query:** select language as  
Languages,round(100\*count(\*)/total,2)AS  
Percentage,sub.total  
from job\_data  
cross join(select count(\*)as total from job\_data)as sub  
group by language,sub.total;

#### Output:

Languages	Percentage	total
English	12.50	8
Arabic	12.50	8
Persian	37.50	8
Hindi	12.50	8
French	12.50	8
italian	12.50	8

#### 4. Duplicate Rows Detection:

Objective: Identify duplicate rows in the data.

**Task:** Write an SQL query to display duplicate rows from the job\_data table.

**Query:** select actor\_id,count(\*) as Duplicates from job\_data group by actor\_id having count(\*)>1;

**Output:**

actor_id	Duplicates
1003	2

## Case Study 2: Investigating Metric Spike

### 1.Weekly User Engagement:

Objective: Measure the activeness of users on a weekly basis.

**Task:** Write an SQL query to calculate the weekly user engagement.

**Query:** select extract(week from occurred\_at) as week\_number,  
count(distinct user\_id)as active\_user  
from events  
where event\_type='engagement'  
group by week\_number  
order by week\_number;

**Output:**

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

<b>32</b>	<b>1225</b>
<b>33</b>	<b>1225</b>
<b>34</b>	<b>1204</b>
<b>35</b>	<b>104</b>

## 2.User Growth Analysis:

Objective: Analyze the growth of users over time for a product.

**Task:** Write an SQL query to calculate the user growth for the product.

**Query:** SELECT extract(MONTH from created\_at) as MonthNum, COUNT(DISTINCT user\_id) AS total\_users  
FROM users  
WHERE activated\_at NOT IN ('')  
group by MonthNum;  
SELECT extract(MONTH from created\_at) as MonthNum,  
COUNT(DISTINCT user\_id) as Users,  
ROUND(((COUNT(DISTINCT  
user\_id)/LAG(COUNT(DISTINCT user\_id), 1)  
OVER (ORDER BY extract(MONTH from created\_at)) -  
1)\*100),2)  
as 'Growth %'FROM users WHERE activated\_at NOT IN ('')  
group by MonthNum;

## Output:

<b>Monthnum</b>	<b>users</b>	<b>Growth%</b>
<b>1</b>	<b>712</b>	
<b>2</b>	<b>685</b>	<b>-3.79</b>
<b>3</b>	<b>765</b>	<b>11.68</b>

<b>4</b>	<b>907</b>	<b>18.56</b>
<b>5</b>	<b>993</b>	<b>9.48</b>
<b>6</b>	<b>1086</b>	<b>9.37</b>
<b>7</b>	<b>1281</b>	<b>17.96</b>
<b>8</b>	<b>1347</b>	<b>5.15</b>
<b>9</b>	<b>330</b>	<b>-75.50</b>
<b>10</b>	<b>390</b>	<b>18.18</b>
<b>11</b>	<b>399</b>	<b>2.31</b>
<b>12</b>	<b>486</b>	<b>21.80</b>

### **3.Weekly Retention Analysis:**

Objective: Analyze the retention of users on a weekly basis after signing up for a product.

**Task:** Write an SQL query to calculate the weekly retention of users based on their sign-up cohort.

**Query:** select extract(week from occurred\_at) as sign\_up\_week,count(distinct user\_id) as 'count of users sign up'  
from events  
where event\_type = 'signup\_flow'  
and event\_name = 'complete\_signup'  
group by sign\_up\_week;

### **Output:**

<b>sign_up_week</b>	<b>count of user signup</b>
<b>17</b>	<b>72</b>
<b>18</b>	<b>163</b>
<b>19</b>	<b>185</b>
<b>20</b>	<b>176</b>

<b>21</b>	<b>183</b>
<b>22</b>	<b>196</b>
<b>23</b>	<b>196</b>
<b>24</b>	<b>229</b>
<b>25</b>	<b>207</b>
<b>26</b>	<b>201</b>
<b>27</b>	<b>222</b>
<b>28</b>	<b>215</b>
<b>29</b>	<b>221</b>
<b>30</b>	<b>238</b>
<b>31</b>	<b>193</b>
<b>32</b>	<b>245</b>
<b>33</b>	<b>261</b>
<b>34</b>	<b>259</b>
<b>35</b>	<b>18</b>

#### **4.Weekly Engagement Per Device:**

Objective: Measure the activeness of users on a weekly basis per device.

**Task:** Write an SQL query to calculate the weekly engagement per device.

**Query:** select  
extract(year from occurred\_at) as year\_num,  
extract(week from occurred\_at) as week\_num,  
device,  
count(distinct user\_id) as no\_of\_users  
from events  
where event\_type = 'engagement'  
group by 1,2,3  
order by 1,2,3;



## Output:

year_num	week_num	device	no_of_users
2014	17	acer aspire desktop	9
2014	17	acer aspire notebook	20
2014	17	amazon fire phone	4
2014	17	asus chromebook	21
2014	17	dell inspiron desktop	18
2014	17	dell inspiron notebook	46
2014	17	hp pavilion desktop	14
2014	17	htc one	16
2014	17	ipad air	27
2014	17	ipad mini	19
2014	17	iphone 4s	21
2014	17	iphone 5	65
2014	17	iphone 5s	42
2014	17	kindle fire	6
2014	17	lenovo thinkpad	86
2014	17	mac mini	6
2014	17	macbook air	54
2014	17	macbook pro	143
2014	17	nexus 10	16
2014	17	nexus 5	40
2014	17	nexus 7	18
2014	17	nokia lumia 635	17
2014	17	samsung galaxy tablet	8
2014	17	samsung galaxy note	7
2014	17	samsung galaxy s4	62

**continue.....(total 491 rows)**

## 5.Email Engagement Analysis:

Objective: Analyze how users are engaging with the email service.

**Task:** Write an SQL query to calculate the email engagement metrics.

**Query:** select

100.0 \* sum(case when email\_cat = 'email\_opened' then 1  
else 0 end)

/sum(case when email\_cat = 'email\_sent' then 1 else 0  
end)

as email\_opening\_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\_clicking\_rate

```
from
(
select *,
case when action in ('sent_weekly_digest',
'sent_reengagement_email')
then 'email_sent'
when action in ('email_open')
then 'email_opened'
when action in ('email_clickthrough')
then 'email_clicked'
end as email_cat
from events
)a;
```

### Output:

**email\_opening\_rate**  
**31.1921**

**email\_clicking\_rate**  
**10.4745**