OPERATION ANALYTICS AND INVESTIGATING METRIC SPIKE

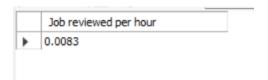
- ➤ **Project Description** Find the valuable insights that can help the company's operations and understand sudden changes in key metrics.
- > Approach -
 - Database creation The database is created, using SQL queries on SQL Workbench.
 - Importing Data Import the data from CSV files into the database, using SQL queries on SQL Workbench.
 - Insights extraction Generating the insights from the database using SQL queries on SQL Workbench.
- > Tech-Stack Used Using SQL Workbench for this project.
- ➤ Insights –

CASE STUDY 1

• Jobs Reviewed Over Time -

```
SELECT COUNT(DISTINCT job_id)/(30*24) as `Job reviewed per hour` FROM job_data WHERE ds BETWEEN '2020-11-01' AND '2020-11-30';
```

The number of jobs reviewed per hour for each day in November 2020.



• Throughput Analysis –

```
SELECT ds as `Date of review`,`Job reviewed`, AVG(`Job reviewed`)

OVER(ORDER BY ds ROWS BETWEEN 6 PRECEDING AND CURRENT ROW) as `Throughput 7 Rolling Average`

FROM ( SELECT ds, COUNT(DISTINCT job_id) as `Job reviewed` From job_data GROUP BY ds ORDER BY ds) a;
```

The 7-day rolling average of throughput.

	Date of review	Job reviewed	Throughput 7 Rolling Average	
•	2020-11-25	1	1.0000	
	2020-11-26	1	1.0000	
	2020-11-27	1	1.0000	
	2020-11-28	2	1.2500	
	2020-11-29	1	1.2000	
	2020-11-30	2	1.3333	

• Language Share Analysis -

```
SELECT distinct language, ((COUNT(language)/(SELECT COUNT(*) FROM job_data))*100) as `Percentage share` FROM job_data group by language;
```

The percentage share of each language over the last 30 days.

	language	Percentage share
•	English	12.5000
	Arabic	12.5000
	Persian	37.5000
	Hindi	12.5000
	French	12.5000
	Italian	12.5000

• Duplicate Rows Detection -

```
SELECT * FROM

( SELECT *, ROW_NUMBER() OVER( PARTITION BY job_id) as `Row Num`

FROM job_data) a WHERE `Row Num` >1;
```

Display duplicate rows from the table.

	job_id	actor_id	event	language	time_spent	org	ds	Row Num
•	23	1005	transfer	Persian	22	D	2020-11-28	2
	23	1004	skip	Persian	56	Α	2020-11-26	3

CASE STUDY 2

• Weekly User Engagement –

```
-- TASK A

SELECT EXTRACT(WEEK FROM occurred_at) AS `Week Number`,

COUNT(DISTINCT user_id) AS `Number of Users`

FROM events GROUP BY `Week Number`;
```

The weekly user engagement.

	Week Number	Number of 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

• User Growth Analysis –

```
-- TASK B

SELECT 'Year Num', 'Week Num', 'Num of Active Users', SUM('Num of Active Users')

OVER(ORDER BY 'Year Num', 'Week Num') AS 'Cummulative Active Users'

FROM (SELECT EXTRACT(Year FROM activated_at) AS 'Year Num',

EXTRACT(WEEK FROM activated_at) AS 'Week Num',

COUNT(DISTINCT user_id) AS 'Num of Active Users'

FROM users WHERE state = 'active' GROUP BY 'Year Num', 'Week Num'

ORDER BY 'Year Num', 'Week Num') a;
```

The user growth for the product.

	Year Num	Week Num	Num of Active Users	Cummulative Active Users
٠	2013	0	23	23
	2013	1	30	53
	2013	2	48	101
	2013	3	36	137
	2013	4	30	167
	2013	5	48	215
	2013	6	38	253
	2013	7	42	295
	2013	8	34	329
	2013	9	43	372
	2013	10	32	404
	2013	11	31	435
	2013	12	33	468
	2013	13	39	507
	2013	14	35	542
	2013	15	43	585
	2013	16	46	631
	2013	17	49	680
	2013	18	44	724
	2013	19	57	781
	2013	20	39	820
	2013	21	49	869
	2013	22	54	923

Year Num	Week Num	Num of Active Users	Cummulative Active Users
2013	23	50	973
2013	24	45	1018
2013	25	57	1075
2013	26	56	1131
2013	27	52	1183
2013	28	72	1255
2013	29	67	1322
2013	30	67	1389
2013	31	67	1456
2013	32	71	1527
2013	33	73	1600
2013	34	78	1678
2013	35	63	1741
2013	36	72	1813
2013	37	85	1898
2013	38	90	1988
2013	39	84	2072
2013	40	87	2159
2013	41	73	2232
2013	42	99	2331
2013	43	89	2420
2013	44	96	2516
2013	45	91	2607

Year Num	Week Num	Num of Active Users	Cummulative Active Users
2013	45	91	2607
2013	46	88	2695
2013	47	102	2797
2013	48	97	2894
2013	49	116	3010
2013	50	124	3134
2013	51	102	3236
2013	52	47	3283
2014	0	83	3366
2014	1	126	3492
2014	2	109	3601
2014	3	113	3714
2014	4	130	3844
2014	5	133	3977
2014	6	135	4112
2014	7	125	4237
2014	8	129	4366
2014	9	133	4499
2014	10	154	4653
2014	11	130	4783
2014	12	148	4931

Year Num	Week Num	Num of Active Users	Cummulative Active Users
2014	13	167	5098
2014	14	162	5260
2014	15	164	5424
2014	16	179	5603
2014	17	170	5773
2014	18	163	5936
2014	19	185	6121
2014	20	176	6297
2014	21	183	6480
2014	22	196	6676
2014	23	196	6872
2014	24	229	7101
2014	25	207	7308
2014	26	201	7509
2014	27	222	7731
2014	28	215	7946
2014	29	221	8167
2014	30	238	8405
2014	31	193	8598
2014	32	245	8843
2014	33	261	9104
2014	34	259	9363
2014	35	18	9381

• Weekly Retention Analysis -

```
-- TASK C

SELECT DISTINCT user_id, COUNT(user_id) as user_num,

SUM(CASE WHEN retention_week = 1 THEN 1 ELSE 0 END) AS Per_week_retention

FROM (SELECT a.user_id, a.signup_week, b.engagement_week,

b.engagement_week - a.signup_week AS retention_week

FROM (SELECT DISTINCT user_id, EXTRACT(WEEK FROM occurred_at) AS signup_week FROM events

WHERE event_type = 'signup_flow' AND event_name = 'complete_sigbup') a LEFT JOIN

(SELECT DISTINCT user_id, EXTRACT(WEEK FROM occurred_at) AS engagement_week FROM events

WHERE event_type = 'engagement' ) b ON a.user_id = b.user_id ) d

GROUP BY user_id ORDER BY user_id;
```

The weekly retention of users based on their sign-up cohort.

▶ 11768 1 0 11770 1 0 11775 2 1 11778 3 0 11779 5 1 11780 2 1 11785 1 0 11787 3 1 11791 2 1 11793 6 1 11795 2 1 11799 10 1 11801 2 1 11804 2 1 11806 1 0 11811 2 1 11813 6 0 11816 3 0		user_id	user_num	Per_week_retention
11775 2 1 11778 3 0 11779 5 1 11780 2 1 11785 1 0 11787 3 1 11791 2 1 11793 6 1 11795 2 1 11798 6 1 11799 10 1 11801 2 1 11804 2 1 11806 1 0 11811 2 1 11813 6 0 11816 3 0	•	11768	1	0
11778 3 0 11779 5 1 11780 2 1 11785 1 0 11787 3 1 11791 2 1 11793 6 1 11795 2 1 11798 6 1 11799 10 1 11801 2 1 11804 2 1 11806 1 0 11819 1 0 11811 2 1 11813 6 0 11816 3 0		11770	1	0
11779 5 1 11780 2 1 11785 1 0 11787 3 1 11791 2 1 11793 6 1 11795 2 1 11798 6 1 11799 10 1 11801 2 1 11804 2 1 11806 1 0 11819 1 0 11811 2 1 11813 6 0 11816 3 0		11775	2	1
11780 2 1 11785 1 0 11787 3 1 11791 2 1 11793 6 1 11795 2 1 11798 6 1 11799 10 1 11801 2 1 11804 2 1 11806 1 0 11819 1 0 11811 2 1 11813 6 0 11816 3 0		11778	3	0
11785 1 0 11787 3 1 11791 2 1 11793 6 1 11795 2 1 11798 6 1 11799 10 1 11801 2 1 11804 2 1 11806 1 0 11819 1 0 11811 2 1 11813 6 0 11816 3 0		11779	5	1
11787 3 1 11791 2 1 11793 6 1 11795 2 1 11798 6 1 11799 10 1 11801 2 1 11804 2 1 11806 1 0 11819 1 0 11811 2 1 11813 6 0 11816 3 0		11780	2	1
11791 2 1 11793 6 1 11795 2 1 11798 6 1 11799 10 1 11801 2 1 11804 2 1 11806 1 0 11819 1 0 11811 2 1 11813 6 0 11816 3 0		11785	1	0
11793 6 1 11795 2 1 11798 6 1 11799 10 1 11801 2 1 11804 2 1 11806 1 0 11809 1 0 11811 2 1 11813 6 0 11816 3 0		11787	3	1
11795 2 1 11798 6 1 11799 10 1 11801 2 1 11804 2 1 11806 1 0 11809 1 0 11811 2 1 11813 6 0 11816 3 0		11791	2	1
11798 6 1 11799 10 1 11801 2 1 11804 2 1 11806 1 0 11809 1 0 11811 2 1 11813 6 0 11816 3 0		11793	6	1
11799 10 1 11801 2 1 11804 2 1 11806 1 0 11809 1 0 11811 2 1 11813 6 0 11816 3 0		11795	2	1
11801 2 1 11804 2 1 11806 1 0 11809 1 0 11811 2 1 11813 6 0 11816 3 0		11798	6	1
11804 2 1 11806 1 0 11809 1 0 11811 2 1 11813 6 0 11816 3 0		11799	10	1
11806		11801	2	1
11809 1 0 11811 2 1 11813 6 0 11816 3 0		11804	2	1
11811 2 1 11813 6 0 11816 3 0		11806	1	0
11813 6 0 11816 3 0		11809	1	0
11816 3 0		11811	2	1
		11813	6	0
		11816	3	0
11818 2 1		11818	2	1
11820 4 1		11820	4	1

There are more than 3000 rows that's why I'm attaching the result sheet <u>link</u>.

• Weekly Engagement Per Device –

```
-- TASK D

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;
```

The weekly engagement per device is-

	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

The result file <u>link</u>.

• Email Engagement Analysis -

```
-- TASK E

SELECT 100*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*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 email_events ) a;
```

The email engagement metrics.

	email_opening_rate	email_clicking_rate
•	33.5834	14.7899

➤ Results –

- Fewer than 0.01 jobs were reviewed per hour throughout the month of November.
- To calculate the throughput, we will use a 7-day rolling average. The 7-day rolling average provides the mean for each day over the previous seven days, from day 1 to day 7. Whereas, the daily metric only gives the average for that specific day.
- The Persian language had the largest share among all other languages.
- User engagement was highest in the 28th week and lowest in the 17th week.
- The 34th week of 2014 had the highest number of active users. User growth increased from 23 in the first week of 2013 to 9,381 by the 35th week of 2014.
- The majority of retained users were only retained for a week, after which retention rates dropped.
- Users with the highest engagement with the product were using Macbook Prodevices.
- The email open rate is 33.58%, and the click rate is 14.78% of all the emails.