OPERATIONAL ANALYTICS AND INVESTIGATING METRIC

PROJECT DESCRIPTION: The objective of the project is to analyse the company's end to end operations which helps identifying the area for improvement. As the Data Analyst of Microsoft, I will work closely with the marketing, operation and support team to derive valuable insights from the collected data source. The project also focuses on investing metric spikes in daily engagement and sales, investing their causes and providing insights to address any dip in these metrices

APPROACH:

- 1. Project involves understanding and gathering data sets and tables from various sources and departments within company
- 2. Making sure that the given variable and attribute can be understood properly
- 3. Importing files into my SQL Workbench and through SQL the collected data will be analysed to give insights and patterns
- 4. Key performance metrices are such as daily engagement, sales and customer satisfaction will be identified in collaboration with cross-functional teams
- 5. Cross checking the queries one or two times so that the code will run without errors
- 6. Once the questions are answered it will be saved with screenshot
- 7. Finally, it will be added to the document

TECH-STACK USED:

8. MY SQL Community server-GPL Version 8.0.29 and connector version c++8.0.29 for creating my project as we know MY SQL community server is an open-source relational database management system that uses SQL



INSIGHTS:

- While making the project I learn about SQL how to implement the queries and about the various built-in functions that can be used to get desired output
- I got good bit of exposure about SQL and the functions of that can be used in analysing data from the dataset and also the queries how they worked and executed.

RESULTS:

CASE STUDY 1: JOB DATA ANALYSIS

A. JOBS REVIEWED OVER TIME:

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

```
select ds as date,
count(job_id) as joint_job_id,
round((sum(time_spent) / 3600), 2) as total_time_spent_hr,
round((count(job_id)/ (sum(time_spent) / 3600)), 2) as "jobs reviewd per hour per day" from job_data
WHERE ds BETWEEN "2020-11-01" AND "2020-11-30" group by ds
order by ds;
```

		The second secon		Export: Wrap Cell Content:				
	date	joint_job_id	total_time_spent_hr	jobs reviewd per hour per day				
•	2020-11-25	3	0.04	80.00				
	2020-11-26	3	0.05	64.29				
	2020-11-27	3	0.09	34.62				
	2020-11-28	6	0.03	218.18				
	2020-11-29	3	0.02	180.00				
	2020-11-30	6	0.03	180.00				

B. THROUGHPUT ANALYSIS:

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.

WEEKLY THROUGHPUT:

```
select round((count(event)/sum(time_spent)),2) as weekly_throughput
from job_data;
```

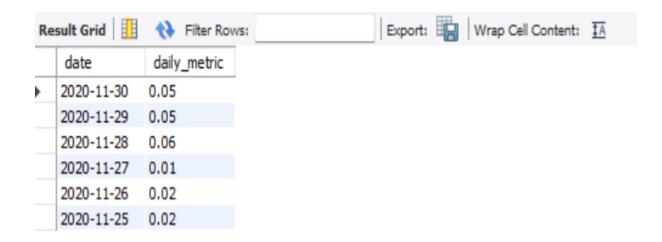
```
weekly_throughput

0.03
```

DAILY THROUGHPUT:

```
# Daily metric throughput
```

select ds as date, round((count(event)/sum(time_spent)),2) as daily_metric
from job data group by date;

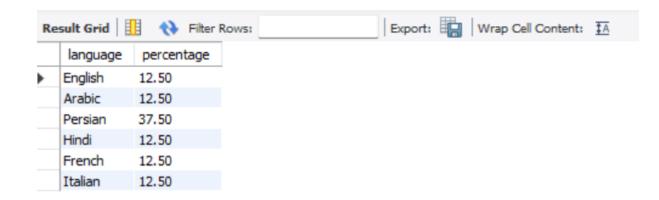


C. LANGUAGE SHARE ANALYSIS:

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

```
select language, round(100 * count(*) / total ,2) as percentage
from job_data
  cross join

(select count(*) as total
  from job_data) as jd
  group by language,jd.total;
```



D. **DUPLICATE ROWS DETECTION**

Task: Write an SQL query to display duplicate rows from the job data table.

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



Case Study 2: INVESTIGATING METRIC SPIKE

1. WEEKLY USER ENGAGEMENT:

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

```
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;
```

	week_number	active_user
•	17	663
	18	1068
	19	1113
	20	1154
	21	1121
	22	1186
	23	1232
	24	1275
	25	1264
	26	1302

week_number	active_user
27	1372
28	1365
29	1376
30	1467
31	1299
32	1225
33	1225
34	1204
35	104

2. USER GROWTH ANALYSIS:

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

```
• 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) as 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;
```

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year	week_number	num_of_users	cumulative_users		
2014	31	193	8598		
2014	32	245	8843		
2014	33	261	9104		
2014	34	259	9363		
2014	35	18	9381		

3. WEEKLY RETENTION ANALYSIS:

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

```
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"
```

```
select
  m.user_id,
  m.login_week,
  n.first,
  m.login_week - n.first as week_number
select
  user_id,
  extract(week from occurred_at) as login_week
  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;
```

Result Grid	Filter R	ows:		Export	: 📳 Wr	ap Cell Con	tent: <u>‡A</u>									
week_numbers	week_0	week_1	week_2	week_3	week_4	week_5	week_6	week_7	week_8	week_9	week_10	week_11	week_12	week_13	week_14	we
17	663	472	324	251	205	187	167	146	145	145	136	131	132	143	116	91
18	596	362	261	203	168	147	144	127	113	122	106	118	127	110	97	85
19	427	284	173	153	114	95	91	81	95	82	68	65	63	42	51	49
20	358	223	165	121	91	72	63	67	63	65	67	41	40	33	40	0
21	317	187	131	91	74	63	75	72	58	48	45	39	35	28	2	0
าา	376	224	150	107	97	72	63	60	55	49	41	30	21	1	n	٥
week_1	.5	wee	k_16	V	veek	_17	W	eek_	18							
91		82		7	7		5									
85	(67		4			0									
49		2		0			0									

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4. WEEKLY ENGAGEMENT PER DEVICE:

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

```
select
extract(week from occurred_at) as week_number,
count(distinct case when device= 'dell inspiration notebook' then user_id else null end ) as dell_inspiration
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= 'ipaid air' then user_id else null end ) as ipaid_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= '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 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 365' 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= 'hd pavilion desktop' then user_id else null end ) as hd_pavilion_desktop,
count(distinct case when device= 'dell inspiron desktop' then user_id else null end ) as dell_inspiron_desktop
from
events
where
event_type='engagement'
group by
week_number
order by
week_number;
```

19 20 21 22 23	0 0 0	115 125 137	44 55	79 79	0	16 21	112 119	36	21	12	87
21 22 23	0			79	0	21	110	22	22		
22 23 24	_	137	AE.				119	32	23	11	103
23 74	0		45	74	0	17	110	23	30	5	91
74		125	45	71	0	15	145	34	21	5	96
	0	152	53	79	0	14	124	33	25	16	88
	n	147	23	70	0 _	าา	157	30	25	11	97
ılt Grid 📙 🖠	Filter Rows:	Ex	port: 📳 V	Vrap Cell Cont	ent: <u>‡A</u>						
week_number	dell_inspiration_notebook	iphone_5	iphone_4s	iphone_5s	ipaid_air	windows_surface	macbook_air	ipad_mini	kindle_fire	amazon_fire_phone	nexus_5
3	0	152	53	79	0	14	124	33	25	16	88
24	0	142	53	79	0	22	152	39	25	11	87
25	0	137	40	78	0	22	121	30	24	13	89
26	0	152	50	94	0	21	134	43	26	13	87
27	0	163	67	83	0	33	142	35	25	10	84
28	0	151	61	03	n	22	149	१६	21	6	95
week_number		<u></u>	-				macbook_air	ipad_mini	kindle_fire	amazon_fire_phone	nexus_5
26	0	152	50	94	0	21	134	43	26	13	87
27	0	163	67	83	0	33	142	35	25 31	10	84
28 29	0	151 144	61	93 90	0	33 28	148 148	35 34	37	12	85 77
30	0	152	65	103	0	19	159	35	25	12	84
31	0	135	56	71	0	10	147	27	14	14	60
ult Grid	National Property of the Prope	Ex	oport: 📳 V	Vrap Cell Cont	ent: <u>‡A</u>						
week_number	dell_inspiration_notebook	iphone_5	iphone_4s	iphone_5s	ipaid_air	windows_surface	macbook_air	ipad_mini	kindle_fire	amazon_fire_phone	nexus_5
	0	135	56	71	0	19	147	27	14	14	69
31	V							20	10	10	67
	0	119	34	67	0	10	125	30	12	12	0/
31	-	119 110	34 35	67 65	0	10 15	125		14	14	70
31 32	0							28	14	14	

5. EMAIL ENGAGEMENT ANALYSIS:

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

```
select
  100.0* sum(case when email_action = 'email_open' then 1 else 0 end) /
  sum(case when email action = 'email sent' then 1 else 0 end) as email open_rate,
  100.0* sum(case when email action = 'email clicked' then 1 else 0 end) /
  sum(case when email_action = 'email_sent' then 1 else 0 end) as email_clicked_rate
  from
⊖ (select *,
when action in ('sent weekly digest', 'sent reengagement email') then 'email sent'
   when action in ('email_open') then 'email_open'
   when action in ('email_clickthrough') then 'email_clicked'
   else null
   end as email action
   from
   project3.email events
    ) a;
                                              Export: Wrap Cell Content: IA
Result Grid 🔢 🙌 Filter Rows:
    email_open_rate | email_clicked_rate
33.58339
                   14.78989
```

THANK YOU