



OPERATION ANALYTICS AND INVESTIGATING METRIC SPIKE

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Project Description

- Operational Analytics, a vital process in business, entails end-to-end operations analysis, enabling Data Analysts to collaborate with diverse teams and extract valuable insights for continuous improvement.
- Operational analytics depends on the skillful investigation of metric spikes, which reveals sudden changes in critical indicators such as a rise in daily user engagement or a fall in sales. It is crucial for a Data Analyst to develop the ability to investigate these variations on a daily basis, which calls for a thorough understanding of the causes of these metric spikes.
- Working as a Lead Data Analyst for a corporation such as Microsoft , I will use SQL knowledge to draw conclusions from a variety of datasets and respond to requests from different departments. The primary goal is to improve operational effectiveness and shed light on the reasons for changes in important indicators so that decisions can be made with knowledge.

Approach



- Importing the Dataset: The first step is maintaining a file path to export data (.csv) into the SQL workbench.
- Understanding the Schema: The next step is to examine the structure of the table holding the data (job_data, events, etc).
- Identifying the Key tables: Identification of the primary key from each of the tables of job_data, email_events, events, users etc.
- Checking for null values: Before the analysis, it is necessary to check for null values in the given tables
- Visually Appealing: The SQL Queries need to be properly formatted so that they can be understood by any user.

CASE STUDY 1: Job Data Analysis

A. Jobs Reviewed Over Time:

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

```
33  -- Task 1: Jobs Reviewed Over Time
34  • SELECT
35      COUNT(job_id) AS number_of_jobs,
36      ROUND(SUM(time_spent) / 3600, 3) AS hours_per_day
37  FROM
38      Job_Data
39  WHERE
40      ds BETWEEN '2020-11-01' AND '2020-11-30'
41  GROUP BY ds;
42
43  -- Task 2: Throughput Analysis
```

Result Grid		
Filter Rows: <input type="text"/>		
Export:  Wrap Cell Content: 		
	number_of_jobs	hours_per_day
▶	2	0.011
	1	0.006
	2	0.009
	1	0.029
	1	0.016
	1	0.013

B. Throughput Analysis:

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

```
43  -- Task 2: Throughput Analysis
44
45 • SELECT
46  AVG(number_of_events) OVER(ROWS BETWEEN 6 PRECEDING AND CURRENT ROW) AS 7_day_rolling_avg
47  FROM
48  (SELECT
49    COUNT(DISTINCT event) AS number_of_events
50  FROM
51    Job_Data
52  GROUP BY ds) AS sub;
```

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

	7_day_rolling_avg
▶	1.0000
	1.0000
	1.0000
	1.2500
	1.2000
	1.3333

C. Language Share Analysis:

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

Persian language has most of the share among other languages in the past 30 days

```
54  -- Task 3: Language Share Analysis
55  •  SELECT
56      language,
57      ROUND((COUNT(language) / (SELECT
58          COUNT(*)
59          FROM
60              Job_Data))) * 100,
61      2) AS lang_share
62  FROM
63      Job_Data
64  WHERE
65      ds > (SELECT
66          MAX(ds) - INTERVAL 30 DAY
67          FROM
68              Job_Data)
69  GROUP BY language;
```

Result Grid

	language	lang_share
▶	English	12.50
	Arabic	12.50
	Persian	37.50
	Hindi	12.50
	French	12.50
	Italian	12.50

D. Duplicate Rows Detection:

Write an SQL query to display duplicate rows from the job_data table.

There are no duplicate rows in the given Job_Data table.
Hence there is no output.

```
71  -- Task 4: Duplicate Rows Detection
72
73 • SELECT
74     job_id, COUNT(*)
75 FROM
76     Job_Data
77 GROUP BY job_id,actor_id,event,language,time_spent,org,ds
78 HAVING COUNT(*) > 1;
79
80
81
82
```

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

job_id	COUNT(*)
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CASE STUDY 2: Investigating Metric Spike

A. Weekly User Engagement:

Write an SQL query to calculate the weekly user engagement.

```
77  -- Task 1: Weekly User Engagement
78 • SELECT
79      EXTRACT(WEEK FROM occurred_at) AS week_num,
80      COUNT(DISTINCT user_id) AS active_users
81  FROM
82      events
83  WHERE
84      event_type = 'engagement'
85  GROUP BY week_num;
86
```

Result Grid			Filter Rows:	Export:	Wrap Cell Content:
	week_num	active_users			
▶	17	663			
	18	1068			
	19	1113			
	20	1154			
	21	1121			
	22	1186			
	23	1232			

B. User Growth Analysis:

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

```
88  -- Task 2: User Growth Analysis
89  • with tab1 as (select extract(month from created_at) as months,
90    extract(year from created_at) as years,
91    count(*) as freq from users
92    group by months,years)
93
94  select years,months,sum(freq) over (order by years,months) as frequency,
95  freq as user_growth from tab1;
96
```

	years	months	frequency	user_growth
▶	2013	1	160	160
	2013	2	320	160
	2013	3	470	150
	2013	4	651	181
	2013	5	865	214
	2013	6	1078	213
	2013	7	1362	284

Result 2 ×

C. Weekly Retention Analysis:

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

```
97  -- Task 3: Weekly Retention Analysis
98 • select * from events;
99 • with retention as(
100     select e.user_id,extract(week from created_at) as weeks_no,
101     min(case when event_type = 'engagement' then extract(week from occurred_at) end) as login_week
102     from users u join events e on u.user_id = e.user_id
103     group by e.user_id,weeks_no),
104     week_retention as (select *,login_week - weeks_no as weeks_retained from retention order by weeks_retained DESC)
105     select weeks_retained, count(user_id) as no_of_users from week_retention
106     group by weeks_retained order by weeks_retained;
107
108
```

Result Grid	Filter Rows:	Export:	Wrap Cell Content:
weeks_retained	no_of_users		
-35	2		
-34	6		
-33	18		
-32	20		
-31	18		
-30	26		
-29	24		

D. Weekly Engagement Per Device:

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

```
109 -- Task 4: Weekly Engagement Per Device
110
111 • SELECT
112     EXTRACT(WEEK FROM occurred_at) AS weeks,
113     device,
114     COUNT(DISTINCT user_id)
115 FROM
116     events
117 WHERE
118     event_type = 'engagement'
119 GROUP BY weeks , device
120 ORDER BY weeks , device;
121
```

Result Grid | Filter Rows: | Export: | Wrap Cell Contents: [1A](#)

	weeks	device	COUNT(DISTINCT user_id)
▶ 17	17	acer aspire desktop	9
	17	acer aspire notebook	20
	17	amazon fire phone	4
	17	asus chromebook	21
	17	dell inspiron desktop	18
	17	dell inspiron notebook	46
	17	hp pavilion desktop	14

Result 4 x

E. Email Engagement Analysis:

Write an SQL query to calculate the email engagement metrics.

```
121
122
123  -- Task 5: Email Engagement Analysis
124 • SELECT
125     EXTRACT(WEEK FROM occurred_at) AS weeks, action, COUNT(*)
126 FROM
127     email_events
128 GROUP BY weeks , action
129 ORDER BY weeks;
130
131
```

Result Grid | Filter Rows: | Export: | Wrap Cell Content: [IA](#)

	weeks	action	COUNT(*)
▶	17	email_clickthrough	166
	17	email_open	310
	17	sent_reengagement_email	73
	17	sent_weekly_digest	908
	18	email_clickthrough	430
	18	email_open	912
	18	sent_reengagement_email	157

Tech-Stack

- **MySQL Workbench:** The main interactive development environment for SQL queries is MySQL Workbench (8.0.34). For data analysis, it makes query creation, execution, and debugging more efficient.
- **Windows Function:** Does calculation across a set of rows that are related to the current row. These functions are used when we want to calculate Average Running Price, Running Total Orders, Running Sum Sales, Rank and Percentile.

Insights

1. Job Data Analysis

In case study 1, we were able to analyze job reviewed per hour for each day in the month of November, 2020. Also some other insights like removal of duplicate rows, percentage share of each language in the past 30 days, and 7-day rolling average throughput.

2. Investigating Metric Spike

In Case Study 2, we were able to understand how different users engage with email events in the application. Were able to analyse growth of users over time for a product, were able to analyse and measure activeness of a user on a weekly basis per device.

Result

- Remembering to adapt these queries on specific database schemas.
- These learned insights helped me understand specific business questions which were addressed by SQL queries.
- Learning about the SQL clauses such as the join clauses and sub-queries. The importance of order by and group by and many more.
- We were able to import the dataset from a .CSV file into the SQL workbench for performing analysis.
- Achieving the ability to learn and write SQL queries to execute different business questions.
- Solving business related problems using Windows functions of SQL.

Thankyou