

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.

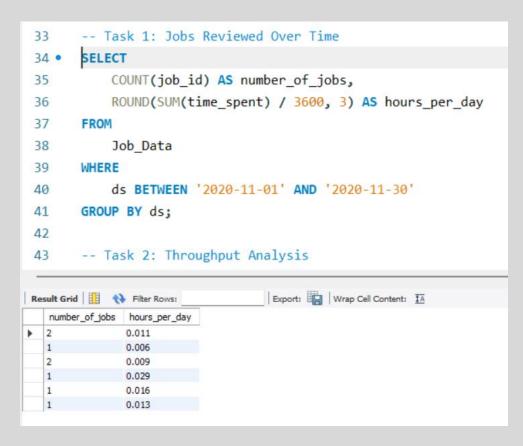
Approach

- 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

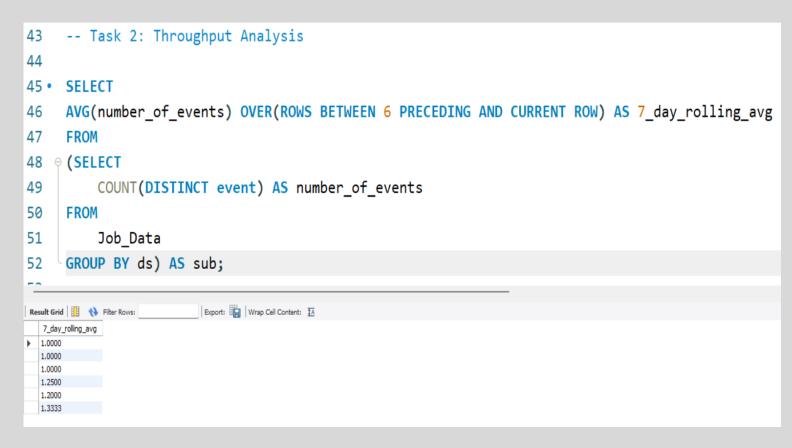
A. Jobs Reviewed Over Time:

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



B. Throughput Analysis:

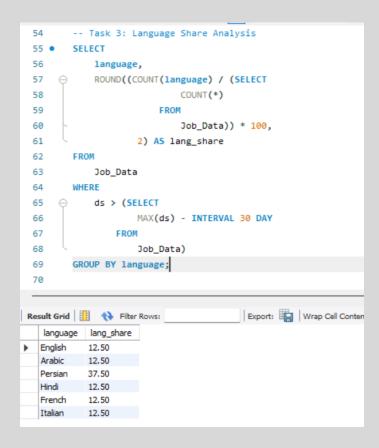
Calculate the 7-day rolling average of throughput (number of events per second).



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



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(*)
      FROM
76
           Job Data
      GROUP BY job_id,actor_id,event,language,time_spent,org,ds
78
      HAVING COUNT(*) > 1;
79
80
81
82
                              Export: Wrap Cell Content: TA
Result Grid Filter Rows:
  job_id COUNT(*)
```

CASE STUDY 2: Investigating Metric Spike

A. Weekly User Engagement:

Write an SQL query to calculate the weekly user engagement.

```
-- Task 1: Weekly User Engagement
78 • SELECT
          EXTRACT(WEEK FROM occoured_at) AS week_num,
          COUNT(DISTINCT user id) AS active users
     FROM
82
          events
     WHERE
          event_type = 'engagement'
     GROUP BY week_num;
86
Result Grid Filter Rows:
                         Export: Wrap Cell Content: TA
  week_num active_users
```

B. User Growth Analysis:

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

```
-- 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,
    count(*) as freq from users
91
    group by months, years)
93
    select years, months, sum(freq) over (order by years, months) as frequency,
94
    freq as user_growth from tab1;
96
                  Export: Wrap Cell Content: IA
```

C. Weekly Retention Analysis:

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

```
-- Task 3: Weekly Retention Analysis
98 • select * from events;
99 • ⊖ with retention as(
      select e.user_id,extract(week from created_at) as weeks_no,
100
      min(case when event_type = 'engagement' then extract(week from occoured_at) end) as login_week
101
      from users u join events e on u.user id = e.user id
102
      group by e.user_id,weeks_no),
103
      week_retention as (select *,login_week - weeks_no as weeks_retained from retention order by weeks_retained DESC)
      select weeks_retained, count(user_id) as no_of_users from week_retention
105
      group by weeks_retained order by weeks_retained;
106
107
108
                         Export: Wrap Cell Content: IA
 Result Grid Filter Rows:
   weeks_retained no_of_users
```

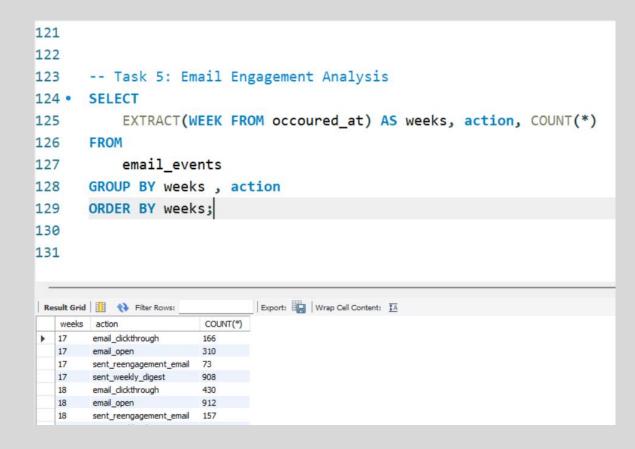
D. Weekly Engagement Per Device:

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

```
-- Task 4: Weekly Engagement Per Device
110
111 • SELECT
            EXTRACT(WEEK FROM occoured at) AS weeks,
112
113
            device,
114
            COUNT(DISTINCT user_id)
115
       FROM
116
            events
       WHERE
117
            event type = 'engagement'
118
       GROUP BY weeks, device
       ORDER BY weeks , device;
121
Export: Wrap Cell Content: IA
                        COUNT (DISTINCT
                        user_id)
        acer aspire desktop
        amazon fire phone
        dell inspiron desktop
        dell inspiron notebook 46
        be equilies dealtes
Result 4 x
```

E. Email Engagement Analysis:

Write an SQL query to calculate the email engagement metrics.



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

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

Thankyou