

Trainity

Task 3 : Operation Analytics and Investigating Metric Spike.

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Case Study 1 : Job Data.

Case Study 2: Investigating metric spike.

Case Study 1: Job Data

Q.1) Write an SQL query to calculate the number of jobs reviewed per hour for each day in November 2020.

Solution: A) Non-Distinct

```
28      #Q.1) Calculate the number of jobs reviewed per hour for each day in Nov 2020
29
30      #A.) Non Distinct
31 •    select
32          ds as date_,
33          (count(job_id))/30*24 as job_rev_pr_hr_day
34      from job_data
35      where ds between '2020-11-01' and '2020-11-30'
36      group by ds;
```

Result Grid	Filter Rows:	Export:	Wrap Cell Content:
date_	job_rev_pr_hr_day		
2020-11-30	1.6000		
2020-11-29	0.8000		
2020-11-28	1.6000		
2020-11-27	0.8000		
2020-11-26	0.8000		

B) Distinct

```
38      #B.) Distinct
39
40 •    select
41          ds as date_,
42          (count(distinct job_id))/30*24 as job_rev_pr_hr_day
43      from job_data
44      where ds between '2020-11-01' and '2020-11-30'
45      group by ds;
```

Result Grid	Filter Rows:	Export:	Wrap Cell Content:
date_	job_rev_pr_hr_day		
2020-11-30	1.6000		
2020-11-29	0.8000		
2020-11-28	1.6000		
2020-11-27	0.8000		
2020-11-26	0.8000		

Q.2) 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.

Solution:

```
47      #Q.2) Objective: Calculate the 7-day rolling average of throughput (number of events per second)
48
49      select ds as date_review ,job_reviewed,avg(job_reviewed)
50      over(order by ds rows between 6 preceding and current row) as _7_day_rolling_avg
51  from (
52      select ds ,count(distinct job_id) as job_reviewed
53      from job_data
54      group by ds order by ds) as a;
```

date_review	job_reviewed	_7_day_rolling_avg
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

Q.3) Write an SQL query to calculate the percentage share of each language over the last 30 days.

Solution:

```
56
57      #Q.3) Objective: Calculate the percentage share of each language in the last 30 days
58
59      select language_,
60      count(language_) as language_count,
61      round((count(language_)*100)/sum(count(language_))over(),2) as percentage_share
62  from job_data
63  where ds between '2020-11-01' and '2020-11-30'
64  group by language_;
```

language_	language_count	percentage_share
English	1	12.50
Arabic	1	12.50
Persian	3	37.50
Hindi	1	12.50
French	1	12.50
Italian	1	12.50

Q.4) Write an SQL query to display duplicate rows from the job_data table.

Solution:

```
66      #Q.4) Write an SQL query to display duplicate rows from the job_data table.
67
68      •      select *
69      from
70      (select * ,row_number()over(partition by language_ order by ds) as row_numbers
71      from job_data) as a
72      where row_numbers>1;
73
74
75
```

Result Grid

Filter Rows:

Export:

Wrap Cell Content:



	ds	job_id	actor_id	event_	language_	time_spent	org	row_numbers
	2020-11-28	23	1005	transfer	Persian	22	D	2
	2020-11-29	23	1003	decision	Persian	20	C	3



Case Study 2: Investigating Metric Spike

Q.1) Write an SQL query to calculate the weekly user engagement.

Solution:

```
79      # Q.1)Measure the activeness of users on a weekly basis.
80
81      •      select
82      extract(week from occurred_at) as week_number,
83      count(distinct user_id) as No_of_users
84      from events
85      group by week_number
86      order by week_number;
87
```

Result Grid |   Filter Rows:

Export:  | Wrap Cell Content: 

	week_number	No_of_users
▶	17	663
	18	1068
	19	1113
	20	1154
	21	1121
	22	1186

Result 7 x

Q.2) Write an SQL query to calculate the user growth for the product.

Solution:

```
88 #Q.2)-: Write an SQL query to calculate the user growth for the product.
89
90 with monthlydata as (
91     SELECT
92     DATE_FORMAT(activated_at,'%Y') as Year_start_date,
93     DATE_FORMAT(activated_at,'%m') AS month_start_date,
94     COUNT(DISTINCT user_id) AS total_users
95     FROM users
96     GROUP BY Year_start_date,month_start_date
97     ORDER BY Year_start_date),
98 growthdata as (
99     select Year_start_date,month_start_date,total_users,
100     lag(total_users) over (order by Year_start_date,month_start_date) as Prev_month_users from monthlydata )
101     select Year_start_date,month_start_date,total_users,Prev_month_users,
102     case when prev_month_users is not null
103     then ((total_users-Prev_month_users)/Prev_month_users)*100
104     else null
105     end as Growth_Rate_Percentage
106     from growthdata
107     order by Year_start_date, month_start_date;
```

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

	Year_start_date	month_start_date	total_users	Prev_month_users	Growth_Rate_Percentage
▶	2013	01	160	NULL	NULL
	2013	02	160	160	0.0000
	2013	03	150	160	-6.2500
	2013	04	181	150	20.6667
	2013	05	214	181	18.2320

Default 40

Q.3) Write an SQL query to calculate the weekly retention of users based on their sign-up cohort.

Solution:

```
110 #Q.3)Write an SQL query to calculate the weekly retention of users based on their sign-up cohort.
111
112 • SELECT distinct user_id,COUNT(user_id),
113     SUM(CASE WHEN retention_week = 1 Then 1 Else 0 END) as per_week_retention
114 FROM
115 ( SELECT a.user_id,a.signup_week,b.engagement_week,
116     b.engagement_week - a.signup_week as retention_week
117 FROM
118 (
119 (SELECT distinct user_id, extract(week from occurred_at) as signup_week from events
120 WHERE event_type = 'signup_flow'and event_name = 'complete_signup'
121 )a
122 LEFT JOIN
123 (SELECT distinct user_id, extract(week from occurred_at) as engagement_week FROM events
124 where event_type = 'engagement'
125 )b
126 on a.user_id = b.user_id
127 ))d
128 group by user_id order by user_id;
```

Result Grid			
	user_id	COUNT(user_id)	per_week_retention
▶	11768	1	0
	11770	1	0
	11775	2	1
	11778	3	0
	11779	5	1
	11780	2	1
	11785	1	0
	11787	3	1

Q.4) Write an SQL query to calculate the weekly engagement per device.

Solution:

```
132 #Q.4)write an SQL query to calculate the weekly engagement per device.
133 • select
134     extract(year from occurred_at) as year_num,
135     extract(week from occurred_at) as week_num,
136     count(distinct user_id) as number_of_users,
137     device
138 from
139     events
140 where event_type='engagement'
141 group by year_num,week_num,device
142 order by year_num;
143
```

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

year_num	week_num	number_of_users	device
2014	17	9	acer aspire desktop
2014	17	20	acer aspire notebook
2014	17	4	amazon fire phone
2014	17	21	asus chromebook
2014	17	18	dell inspiron desktop
2014	17	46	dell inspiron notebook

Q.5) Write an SQL query to calculate the email engagement metrics.

Solution:

```
144 #Q.5 Write an SQL query to calculate the email engagement metrics.
145 • SELECT
146     100.0*SUM(CASE when email_category = 'email_opened' then 1 else 0 end)/SUM(CASE when
147     email_category = 'email_sent' then 1 else 0 end) as email_opening_rate,
148     100.0*SUM(CASE when email_category = 'email_clicked' then 1 else 0 end)/SUM(CASE when
149     email_category = 'email_sent' then 1 else 0 end) as email_clicking_rate
150 from
151     (select *,
152     case
153         when action in ('sent_weekly_digest','sent_reengagement_email') then 'email_sent'
154         when action in ('email_open') then 'email_opened'
155         when action in ('email_clickthrough') then 'email_clicked'
156     end as email_category from email)a;
```

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

email_opening_rate	email_clicking_rate
33.58339	14.78989

Software Used: MySQL Workbench 8.0

Thank You.