

Operation Analytics and Investigating Metric Spike

Project Description

This is a detailed report for operation analytics and investigation metric spike for given two operation case studies, deriving certain insights from it.

Following are the operations, to whom we have to provide a detailed report:

Case Study 1 (Job Data)

1. **Number of jobs reviewed:** Amount of jobs reviewed over time.
Your task: Calculate the number of jobs reviewed per hour per day for November 2020?
2. **Throughput:** It is the no. of events happening per second.
Your task: Let's say the above metric is called throughput. Calculate 7 days rolling average of throughput? For throughput, do you prefer daily metric or 7-day rolling, and why?
3. **Percentage share of each language:** Share of each language for different contents.
Your task: Calculate the percentage share of each language in the last 30 days?
4. **Duplicate rows:** Rows that have the same value present in them.
Your task: Let's say you see some duplicate rows in the data. How will you display duplicates from the table?

Case Study 2 (Investigating metric spike on Users, Emails, and Email_Events table):

1. **User Engagement:** To measure the activeness of a user. Measuring if the user finds quality in a product/service.
Your task: Calculate the weekly user engagement?
2. **User Growth:** Amount of users growing over time for a product.
Your task: Calculate the user growth for product?
3. **Weekly Retention:** Users getting retained weekly after signing-up for a product.
Your task: Calculate the weekly retention of users-sign up cohort?
4. **Weekly Engagement:** To measure the activeness of a user. Measuring if the user finds quality in a product/service weekly.
Your task: Calculate the weekly engagement per device?
5. **Email Engagement:** Users engaging with the email service.
Your task: Calculate the email engagement metrics?

Approach

Since the dataset provided is in CSV format, hence here we can freely explore the dataset via different analysis tools to get insights and a better understanding of its schema.

I selected MSSMS for understanding the structure of the tables and how they are connected with each other.

Tech-Stack Used

The software used for the analysis is *Microsoft SQL Server Management Studio 18*.

MSSMS is quite easy to execute the queries. Apart from this, it is easy to learn and beginner friendly, making it my first choice for analysis. Also, its recommendation system comes a lot of handy while writing complex queries.

Insights

Following are all the insights that we got from the dataset after executing different queries.

Case Study 1 (Job Data)

1. **Number of jobs reviewed:** Amount of jobs reviewed over time.

Your task: Calculate the number of jobs reviewed per hour per day for November 2020?

```
select distinct CONVERT(date, ds) as 'Date', COUNT(job_id) as 'Daily Job View',  
round(sum(time_spent)/(60*60),3) as 'Hourly Job View' from job_data  
where MONTH(ds) = 11  
group by ds
```

	Date	Daily Job View	Hourly Job View
1	2020-11-25	1	0.013
2	2020-11-26	1	0.016
3	2020-11-27	1	0.029
4	2020-11-28	2	0.009
5	2020-11-29	1	0.006
6	2020-11-30	2	0.011

2. **Throughput:** It is the no. of events happening per second.

Your task: Let's say the above metric is called throughput. Calculate 7 day rolling average of throughput? For throughput, do you prefer daily metric or 7-day rolling, and why?

(**NOTE:** Since the dataset contains days less than a weekday (i.e 7), hence I have calculated the throughput on Daily Metric).

```
select *, round(avg(Throughput) over(order by day),3) AS 'Rolling Average'
from (
  select *, round([daily events]/[daily time (sec)],3) as 'Throughput'
  from (
    select day(ds) as day,
    count(event) as 'daily events',sum(time_spent) as 'daily time (sec)'
    from job_data
    where MONTH(ds)=11
    group by day(ds)
  )table_1
)table_2
```

	day	daily events	daily time (sec)	Throughput	Rolling Average
1	25	1	45	0.022	0.022
2	26	1	56	0.018	0.02
3	27	1	104	0.01	0.017
4	28	2	33	0.061	0.028
5	29	1	20	0.05	0.032
6	30	2	40	0.05	0.035

3. **Percentage share of each language:** Share of each language for different contents.

Your task: Calculate the percentage share of each language in the last 30 days?

```
select distinct language,SUM(time_spent) as 'Total Tlme',
round((SUM(time_spent)) / (select SUM(time_spent) from job_data),3)*100 as 'Language %'
from job_data
where ds between (select CONVERT(date, (select DATEADD(DAY,-30, (select MAX(ds) from
job_data)))))) and
(select CONVERT(date, (select MAX(ds) from job_data)))
group by language;
```

	language	Total Tlme	Language %
1	Arabic	25	8.4
2	English	15	5
3	French	104	34.9
4	Hindi	11	3.7
5	Italian	45	15.1
6	Persian	98	32.9

4. **Duplicate rows:** Rows that have the same value present in them.

Your task: Let's say you see some duplicate rows in the data. How will you display duplicates from the table?

```
select ROW_NUMBER() over (Partition by CONCAT( actor_id,' ',time_spent) order by CONCAT(
actor_id,' ',time_spent)) as 'Unique_Row_Count' ,
CONCAT( actor_id,' ',time_spent) as 'Unique_id',*
from job_data
```

	Unique_Row_Count	Unique_id	ds	job_id	actor_id	event	language	time_spent	org
1	1	1001 15	2020-11-30 00:00:00.000	21	1001	skip	English	15	A
2	1	1002 11	2020-11-28 00:00:00.000	25	1002	decision	Hindi	11	B
3	1	1003 20	2020-11-29 00:00:00.000	23	1003	decision	Persian	20	C
4	1	1003 45	2020-11-25 00:00:00.000	20	1003	transfer	Italian	45	C
5	1	1004 56	2020-11-26 00:00:00.000	23	1004	skip	Persian	56	A
6	1	1005 22	2020-11-28 00:00:00.000	23	1005	transfer	Persian	22	D
7	1	1006 25	2020-11-30 00:00:00.000	22	1006	transfer	Arabic	25	B
8	1	1007 104	2020-11-27 00:00:00.000	11	1007	decision	French	104	D

Case Study 2 (Investigating metric spike on Users, Emails, and Email_Events table):

1. **User Engagement:** To measure the activeness of a user. Measuring if the user finds quality in a product/service.

Your task: Calculate the weekly user engagement?

```
select *,
sum([weekly_engaging_user])over(order by week) as 'Cum user Engagement',
round( cast( ([weekly_engaging_user]*100.0) /
(select count(distinct user_id) from events where event_type='engagement')
as varchar(30)) ,3) as '% User Engagement'
from (
select year(occurred_at) as year ,month(occurred_at) as month
,datepart(week,occurred_at) as week,
count(distinct user_id) as weekly_engaging_user
from events
where event_type='engagement'
group by year(occurred_at),month(occurred_at),datepart(week,occurred_at)
)table_1
Order by week;
```

	year	month	week	weekly_engaging_user	Cum user Engagement	% User Engagement
1	2014	5	18	663	663	10.795
2	2014	5	19	1068	1731	17.388
3	2014	5	20	1113	2844	18.121
4	2014	5	21	1154	3998	18.789
5	2014	5	22	1121	5119	18.251
6	2014	6	23	1186	6305	19.31
7	2014	6	24	1232	7537	20.059
8	2014	6	25	1275	8812	20.759
9	2014	6	26	1264	10076	20.58
10	2014	7	27	1131	11623	18.414
11	2014	6	27	416	11623	6.773
12	2014	7	28	1372	12995	22.338
13	2014	7	29	1365	14360	22.224
14	2014	7	30	1376	15736	22.403
15	2014	8	31	585	17550	9.525
16	2014	7	31	1229	17550	20.01
17	2014	8	32	1299	18849	21.149
18	2014	8	33	1225	20074	19.945
19	2014	8	34	1225	21299	19.945
20	2014	8	35	1204	22503	19.603
21	2014	8	36	104	22607	1.693

2. **User Growth:** Amount of users growing over time for a product.

Your task: Calculate the user growth for the product?

```
select *, SUM([New User]) over(order by year,month) as 'Cum_New_User'
from (
select YEAR(created_at) as year , month(created_at) as month, count(user_id) as 'New User',
round( cast( (count(created_at)*100.)/(select count(user_id)from users) as varchar(30)),2) as
'% New User'
from users
group by YEAR(created_at),month(created_at)
) table_1
order by year, month;
```

	year	month	New User	% New User	Cum_New_User
1	2013	1	332	1.74	332
2	2013	2	328	1.72	660
3	2013	3	383	2.01	1043
4	2013	4	410	2.15	1453
5	2013	5	486	2.55	1939
6	2013	6	485	2.54	2424
7	2013	7	608	3.19	3032
8	2013	8	636	3.34	3668
9	2013	9	699	3.67	4367
10	2013	10	826	4.33	5193
11	2013	11	816	4.28	6009
12	2013	12	972	5.1	6981
13	2014	1	1083	5.68	8064
14	2014	2	1054	5.53	9118
15	2014	3	1231	6.46	10349
16	2014	4	1419	7.44	11768
17	2014	5	1597	8.38	13365
18	2014	6	1728	9.06	15093
19	2014	7	1983	10.4	17076
20	2014	8	1990	10.44	19066

3. **Weekly Retention:** Users getting retained weekly after signing-up for a product.
Your task: Calculate the weekly retention of users sign up cohort?

```
select COUNT(*) as 'Total Users' from users
```

	Total Users
1	19066

```
select count(distinct user_id) as 'Total Engaging Users' from events  
where event_type='engagement'
```

	Total Engaging Users
1	6142

```
select COUNT(*) as 'User Retained within Week'  
from(  
    select A.user_id, DATEADD(DAY,7,convert(date,created_at)) as '1 week later',  
    min(convert(date, occurred_at)) as occur_date  
    from users A  
    right join  events B  
    on A.user_id=B.user_id  
    where event_type like 'engagement'  
    group by A.user_id, DATEADD(DAY,7,convert(date,created_at)) ,B.user_id  
    having  DATEADD(DAY,7,convert(date,created_at)) < min(convert(date, occurred_at))  
)table1
```

	User Retained within Week
1	2368

```

select(
  (
    (select COUNT(*) as 'User Retained within Week' from(
      select A.user_id, DATEADD(DAY,7,convert(date,created_at)) as '1 week later',
      min(convert(date, occurred_at)) as occur_date
      from users A
      right join  events B
      on A.user_id=B.user_id
      where event_type like 'engagement'
      group by A.user_id, DATEADD(DAY,7,convert(date,created_at)) ,B.user_id
      having DATEADD(DAY,7,convert(date,created_at)) < min(convert(date, occurred_at))
      --order by A.user_id
    )temp
    )*100.0)
  / (select COUNT(*)as 'Total Users' from users )
) as '% user Retained Within week'

```

	% user Retained Within week
1	12.420014685828

4. **Weekly Engagement:** To measure the activeness of a user. Measuring if the user finds quality in a product/service weekly.

Your task: Calculate the weekly engagement per device?

```

select YEAR(occurred_at) as 'Year',Month(occurred_at) as
'Month',DATEPART(WEEK,occurred_at) as week_num,device,
COUNT(user_id)as 'total_user',
COUNT(distinct user_id)as 'unique_user'
from events
where event_type='engagement'
group by Year(occurred_at),month(occurred_at),DATEPART(WEEK,occurred_at), device
order by week_num,total_user desc

```


	Year	Month	week_num	device	total_user_engagement	unique_user
1	2014	5	18	macbook pro	1516	143
2	2014	5	18	lenovo thinkpad	793	86
3	2014	5	18	iphone 5	706	65
4	2014	5	18	dell inspiron notebook	503	46
5	2014	5	18	macbook air	490	54
6	2014	5	18	iphone 5s	473	42
7	2014	5	18	samsung galaxy s4	449	52
8	2014	5	18	nexus 5	382	40
9	2014	5	18	ipad air	330	27
10	2014	5	18	asus chromebook	251	21
11	2014	5	18	iphone 4s	217	21
12	2014	5	18	acer aspire notebook	206	20
13	2014	5	18	ipad mini	205	19
14	2014	5	18	htc one	190	16
15	2014	5	18	dell inspiron desktop	187	18
16	2014	5	18	nexus 7	177	18
17	2014	5	18	nexus 10	145	16
18	2014	5	18	hp pavilion desktop	132	14
19	2014	5	18	nokia lumia 635	128	17
20	2014	5	18	samsung galaxy note	116	7
21	2014	5	18	windows surface	87	10
22	2014	5	18	amazon fire phone	83	4
23	2014	5	18	samsung galaxy tab...	70	8
24	2014	5	18	acer aspire desktop	67	9

Results Messages						
	Year	Month	week_num	device	total_user_engagement	unique_user
25	2014	5	18	mac mini	59	6
26	2014	5	18	kindle fire	57	6
27	2014	5	19	macbook pro	3301	252
28	2014	5	19	lenovo thinkpad	1732	153
29	2014	5	19	macbook air	1604	121
30	2014	5	19	iphone 5	1328	113
31	2014	5	19	samsung galaxy s4	1130	82
32	2014	5	19	dell inspiron notebook	953	77
33	2014	5	19	nexus 5	938	73
34	2014	5	19	iphone 5s	778	73
35	2014	5	19	dell inspiron desktop	683	58
36	2014	5	19	asus chromebook	523	42
37	2014	5	19	ipad air	520	52
38	2014	5	19	iphone 4s	448	46
39	2014	5	19	hp pavilion desktop	373	37
40	2014	5	19	nexus 10	370	30
41	2014	5	19	acer aspire notebook	363	33
42	2014	5	19	nokia lumia 635	341	33
43	2014	5	19	ipad mini	309	30
44	2014	5	19	acer aspire desktop	295	26
45	2014	5	19	kindle fire	265	27
46	2014	5	19	nexus 7	252	30
47	2014	5	19	amazon fire phone	177	9
48	2014	5	19	htc one	174	19

(...continuing to 543 rows. The last window picture is attached below)

	Year	Month	week_num	device	total_user_engagement	unique_user
521	2014	8	36	lenovo thinkpad	123	16
522	2014	8	36	macbook pro	122	17
523	2014	8	36	dell inspiron notebook	66	9
524	2014	8	36	macbook air	64	10
525	2014	8	36	iphone 4s	57	6
526	2014	8	36	asus chromebook	38	6
527	2014	8	36	nexus 5	34	4
528	2014	8	36	kindle fire	32	3
529	2014	8	36	windows surface	30	3
530	2014	8	36	samsung galaxy s4	29	6
531	2014	8	36	acer aspire notebook	28	3
532	2014	8	36	mac mini	25	2
533	2014	8	36	iphone 5s	22	3
534	2014	8	36	ipad mini	21	2
535	2014	8	36	htc one	18	2
536	2014	8	36	nexus 7	17	2
537	2014	8	36	nexus 10	15	2
538	2014	8	36	hp pavilion desktop	10	1
539	2014	8	36	iphone 5	9	2
540	2014	8	36	nokia lumia 635	7	2
541	2014	8	36	acer aspire desktop	7	1
542	2014	8	36	samsung galaxy note	6	1
543	2014	8	36	dell inspiron desktop	4	1

5. **Email Engagement:** Users engaging with the email service.

Your task: Calculate the email engagement metrics?

```
select action,Email_cat as 'User Engagement',
round( cast(((Email_cat*100.0)/
(select COUNT( action ) from email_events where action in
('sent_reengagement_email','sent_weekly_digest')) )as varchar(20)),3) as '% User
Engagement'
from (
    select action,COUNT(action)as Email_cat from email_events
    where action in ('email_clickthrough','email_open')
    group by action
)temp
group by Email_cat,action;
```

	action	User Engagement	% User Engagement
1	email_clickthrough	9010	14.79
2	email_open	20459	33.583

Result:

By completing the project, I am feeling a lot more confident in my SQL knowledge. It really helped me to brush up on my concepts related to Window Functions, Sub-queries, Joins, Views, and Aggregate functions.

It also helped me to understand how much important Operation Analytics and Metrics insights have for any company.

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Task: 3