Case Study 2(Investigating metric spike)

Project Description

In this project we are trying to analyse the given dataset and draw key insights from it. I am given data for users, events, email_events data.

Users table tells us about descriptive information about that user's account like userid, when the user was created for the first time, when the user was activated, current state of the user i.e activated or inactive, the language the user speaks.

Events table tells us about descriptive information about an event where an event is an action that a user has taken. This tells us which user has taken what type of event, event name, when he has performed the action and the device from which user has performed the action. This also gives the location from which the action was performed. These events include login events, messaging events, search events, events logged as users progress through a signup funnel, events around received emails.

Email_Events table tells us about descriptive information about events related to emails only. Approach

The csv datas were given. I have created the tables. Then I have loaded data from the csv files.

```
13 ● ⊖ CREATE TABLE users (
             user_id int ,
14
             created at varchar(60),
15
16
             company_id int,
             language VARCHAR(30),
17
             activated at varchar(60) ,
18
             state VARCHAR(30)
19
20
        );
         LOAD DATA INFILE 'E:\\TrainityAssignments\\users.csv'
21 •
22
        INTO TABLE users
        FIELDS TERMINATED BY '.'
23
         ENCLOSED BY """
        LINES TERMINATED BY '\n'
25
         IGNORE 1 LINES;
26
27
49 • ⊖ CREATE TABLE `events` (
          user_id INT,
50
          occurred_at varchar(60),
51
          event_type VARCHAR(60),
52
53
          event name VARCHAR(60),
54
          location VARCHAR(50),
          device VARCHAR(50),
55
          user_type varchar(10)
56
      );
57
58
59
60 •
      LOAD DATA INFILE 'E:\\TrainityAssignments\\events.csv'
      INTO TABLE 'events'
61
      FIELDS TERMINATED BY ','
62
      ENCLOSED BY """
63
      LINES TERMINATED BY '\n'
64
65
      IGNORE 1 LINES;
```

```
34 • ⊝ CREATE TABLE email events (
35
           user id INT,
36
           occurred_at varchar(60),
           `action` VARCHAR(60),
37
38
           user type int
39
       );
40
41
42 •
       LOAD DATA INFILE 'E:\\TrainityAssignments\\email events.csv'
       INTO TABLE email events
43
       FIELDS TERMINATED BY ','
44
       ENCLOSED BY '"'
45
       LINES TERMINATED BY '\n'
46
47
       IGNORE 1 LINES;
```

I was having some issues loading the fields related to date fields so I had to take date fields as varchar same goes for user_type.

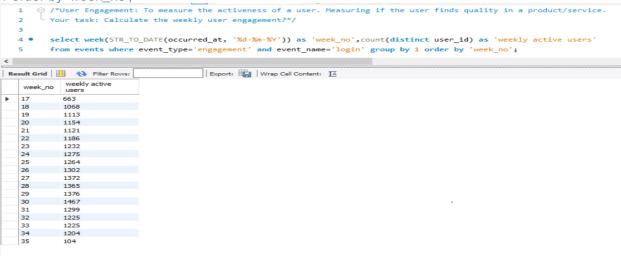
Tech-Stack Used

MySQL Server as the database, MySQL WORKBENCH as a query editor. G-Drive to upload the assignment.

Insights

1. User Engagement: To measure the activeness of a user. Measuring if the userfinds quality in a product/service. We have to calculate the weekly user engagement.

select week(STR_TO_DATE(occurred_at, '%d-%m-%Y')) as 'week_no',count(distinct user_id) as 'weekly active users' from events where event_type='engagement' and event_name='login' group by 1 order by 'week no';



2. User Growth: Number of users growing over time for a product. We have to calculate the user growth for product. – Here I am calculating growth per month

select month(STR_TO_DATE(created_at, '%Y-%m-%d')) as 'month_no',count(user_id) as 'newly created users'

```
from users where state like '%active%' group by 1 order by 1;
  10
          /*User Growth: Amount of users growing over time for a product.
  11
         Your task: Calculate the user growth for product?*/
  12
          select month(STR_TO_DATE(created_at, '%Y-%m-%d')) as 'month_no',count(user_id) as 'newly created users'
  13 •
          from users where state like '%active%' group by 1 order by 1;
  14
  15
 <
 Export: Wrap Cell Content: TA
             newly created
     month_no
    1
             712
    2
             685
    4
             907
             993
    6
             1086
             1281
             1347
             330
    10
             390
    11
             399
    12
             486
```

3. Weekly Retention: Users getting retained weekly after signing-up for a product. We have to calculate the weekly retention of users-sign up cohort.

```
with cte1 as(
select u.user id.e.occurred at as occurred at.u.activated at.
        week(str to date(u.activated at,'%Y-%m-%d')) AS activation week,
        datediff(str to date(e.occurred at, '%d-%m-%Y'), str to date(u.activated at, '%Y-%m-
%d')) AS age at event,
        datediff('2014-09-01 00:00', str to date(u.activated at, '%Y-%m-%d')) AS user age
     FROM users u
     JOIN events e
      ON e.user id = u.user id
      AND e.event_type = 'engagement'
      AND e.event name = 'login'
      #AND e.occurred at >= '01-05-2014'
      #AND e.occurred at < '01-09-2014'
     WHERE u.activated at <>"
     order by user_id)
SELECT week(str to date(cte1.occurred at, '%d-%m-%Y')) AS "week",
    #AVG(cte1.age_at_event) AS "Average age during week",
    COUNT(DISTINCT CASE WHEN cte1.user_age > 77 THEN cte1.user_id ELSE NULL END)
AS "11+ weeks",
    COUNT(DISTINCT CASE WHEN cte1.user_age < 77 AND cte1.user_age >= 70 THEN
cte1.user id ELSE NULL END) AS "10 weeks",
    COUNT(DISTINCT CASE WHEN cte1.user age < 70 AND cte1.user age >= 63 THEN
cte1.user id ELSE NULL END) AS "9 weeks",
```

COUNT(DISTINCT CASE WHEN cte1.user_age < 63 AND cte1.user_age >= 56 THEN cte1.user id ELSE NULL END) AS "8 weeks",

COUNT(DISTINCT CASE WHEN cte1.user_age < 56 AND cte1.user_age >= 49 THEN cte1.user_id ELSE NULL END) AS "7 weeks",

COUNT(DISTINCT CASE WHEN cte1.user_age < 49 AND cte1.user_age >= 42 THEN cte1.user_id ELSE NULL END) AS "6 weeks",

COUNT(DISTINCT CASE WHEN cte1.user_age < 42 AND cte1.user_age >= 35 THEN cte1.user_id ELSE NULL END) AS "5 weeks",

COUNT(DISTINCT CASE WHEN cte1.user_age < 35 AND cte1.user_age >= 28 THEN cte1.user_id ELSE NULL END) AS "4 weeks",

COUNT(DISTINCT CASE WHEN cte1.user_age < 28 AND cte1.user_age >= 21 THEN cte1.user_id ELSE NULL END) AS "3 weeks",

COUNT(DISTINCT CASE WHEN cte1.user_age < 21 AND cte1.user_age >= 14 THEN cte1.user_id ELSE NULL END) AS "2 weeks",

COUNT(DISTINCT CASE WHEN cte1.user_age < 14 AND cte1.user_age >= 7 THEN cte1.user_id ELSE NULL END) AS "1 week",

COUNT(DISTINCT CASE WHEN cte1.user_age < 7 THEN cte1.user_id ELSE NULL END)
AS "Less than a week"

from cte1 group by 1 order by 1

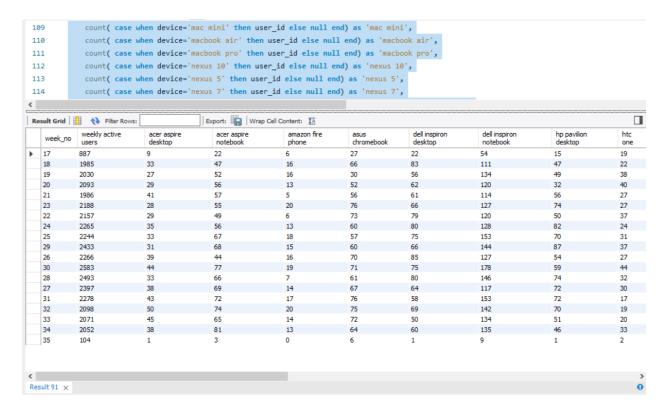
rd	er by	/ 1													
4	3	COUNT(DISTINCT CASE WHEN ctel.user_age < 35 AND ctel.user_age >= 28 THEN ctel.user_id ELSE NULL END) AS "4 weeks",													
44 45 46 47		COUNT(DISTINCT CASE WHEN ctel.user_age < 28 AND ctel.user_age >= 21 THEN ctel.user_id ELSE NULL END) AS "3 weeks",													
		СО	COUNT(DISTINCT CASE WHEN ctel.user_age < 21 AND ctel.user_age >= 14 THEN ctel.user_id ELSE NULL END) AS "2 weeks",												
			COUNT(DISTINCT CASE WHEN ctel.user age < 14 AND ctel.user age >= 7 THEN ctel.user id ELSE NULL END) AS "1 week",												
			`			_			_	_		_		,	
		COUNT(DISTINCT CASE WHEN ctel.user_age < 7 THEN ctel.user_id ELSE NULL END) AS "Less than a week"													
4	8	from cte1													
<															
Re	sult Grid	id III Filter Rows:			Export:	Export: 📳 Wrap Cell Content: 🏗									
	week	11+ weeks	10 weeks	9 weeks	8 weeks	7 weeks	6 weeks	5 weeks	4 weeks	3 weeks	2 weeks	1 week	Less than a week		
•	17	663	0	0	0	0	0	0	0	0	0	0	0	•	
	18	1068	0	0	0	0	0	0	0	0	0	0	0		
	19	1113	0	0	0	0	0	0	0	0	0	0	0		
	20	1154	0	0	0	0	0	0	0	0	0	0	0		
	21	1121	0	0	0	0	0	0	0	0	0	0	0		
	22	1186	0	0	0	0	0	0	0	0	0	0	0		
	23	1232	0	0	0	0	0	0	0	0	0	0	0		
	24	1060	174	0	0	0	0	0	0	0	0	0	0		
	25	915	177	157	0	0	0	0	0	0	0	0	0		
	26	851	102	174	156	0	0	0	0	0	0	0	0		
	27	859	69	99	171	165	0	0	0	0	0	0	0		
	28	804	44	68	98	173	171	0	0	0	0	0	0		
	29	770	33	44	73	96	180	175	0	0	0	0	0		
	30	796	28	33	55	79	99	193	179	0	0	0	0		
	31	666	21	26	42	50	58	95	179	155	0	0	0		
	32	558	12	27	34	36	39	65	87	169	194	0	0		
	33	514	15	17	33	22	19	50	60	81	210	200	0		
	34	464	13	17	21	20	24	30	44	54	106	215	196		
	35	22	0	0	0	0	1	3	1	1	10	6	60		

 Weekly Engagement: To measure the activeness of a user. Measuring if the userfinds quality in a product/service weekly. We have to calculate the weeklyengagement per device.

```
select
       week(STR TO DATE(occurred at, '%d-%m-%Y')) as 'week no',
       count( user id) as 'weekly active users',
  count( case when device='acer aspire desktop' then user id else null end) as 'acer aspire
  count( case when device='acer aspire notebook' then user id else null end) as 'acer aspire
notebook'.
  count( case when device='amazon fire phone' then user id else null end) as 'amazon fire phone',
  count( case when device='asus chromebook' then user_id else null end) as 'asus chromebook'.
  count( case when device='dell inspiron desktop' then user_id else null end) as 'dell inspiron
desktop'.
  count( case when device='dell inspiron notebook' then user_id else null end) as 'dell inspiron
notebook'.
  count( case when device='hp pavilion desktop' then user id else null end) as 'hp pavilion desktop',
  count( case when device='htc one' then user_id else null end) as 'htc one',
  count( case when device='ipad air' then user id else null end) as 'ipad air',
  count( case when device='ipad mini' then user id else null end) as 'ipad mini',
  count( case when device='iphone 4s' then user id else null end) as 'iphone 4s',
  count( case when device='iphone 5' then user id else null end) as 'iphone 5',
  count( case when device='iphone 5s' then user id else null end) as 'iphone 5s',
  count( case when device='kindle fire' then user id else null end) as 'kindle fire',
  count( case when device='lenovo thinkpad' then user_id else null end) as 'lenovo thinkpad',
  count( case when device='mac mini' then user_id else null end) as 'mac mini',
  count( case when device='macbook air' then user id else null end) as 'macbook air',
  count( case when device='macbook pro' then user_id else null end) as 'macbook pro',
  count( case when device='nexus 10' then user id else null end) as 'nexus 10',
  count( case when device='nexus 5' then user id else null end) as 'nexus 5',
  count( case when device='nexus 7' then user id else null end) as 'nexus 7',
  count( case when device='nokia lumia 635' then user id else null end) as 'nokia lumia 635',
  count( case when device='samsumg galaxy tablet' then user id else null end) as 'samsumg
galaxy tablet',
  count( case when device='samsung galaxy note' then user_id else null end) as 'samsung galaxy
note',
  count( case when device='samsung galaxy s4' then user_id else null end) as 'samsung galaxy s4',
  count( case when device='windows surface' then user id else null end) as 'windows surface'
from events where event type='engagement'
and event name='login'
```

group by 1

order by 'week no';



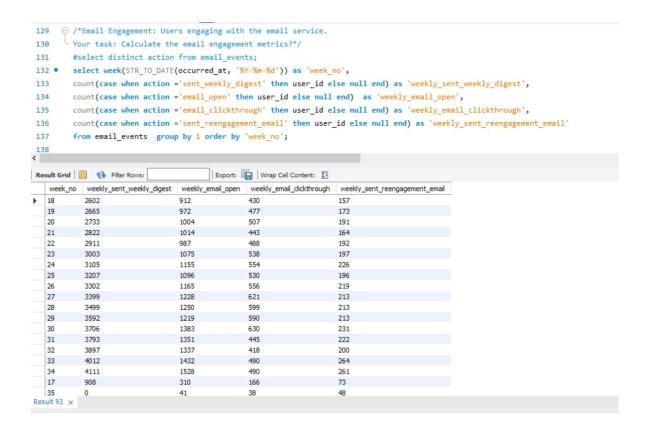
Full Results: Results



5. Email Engagement: Users engaging with the email service. We have to calculate the email engagement metrics.

#select distinct action from email events;

```
select week(STR_TO_DATE(occurred_at, '%Y-%m-%d')) as 'week_no', count(case when action ='sent_weekly_digest' then user_id else null end) as 'weekly_sent_weekly_digest', count(case when action ='email_open' then user_id else null end) as 'weekly_email_open', count(case when action ='email_clickthrough' then user_id else null end) as 'weekly_email_clickthrough', count(case when action ='sent_reengagement_email' then user_id else null end) as 'weekly_sent_reengagement_email' from email_events group by 1 order by 'week_no';
```



Result:

- I was able to create meaningful insights from the dataset which can be crucial for business. I
 was able to sharpen my SQL skills from here. Successfully completing this project boosted
 my confidence.
- 2. Created the database from scratch, loading huge data (above 3lakhs) boosted my confidence
- Wrote many complex queries. Faced many issues like I was not able to load the date fields so had to change the data type and load data then had to convert the string to date filed in each time