## **Database Design**

## 1. Database Implementation

```
*FROM THIS STAGE WE RENAME TABLE "ILLNESSES" TO "CONDITIONS" FOR CONSISTENCY, SORRY FOR THE CONFUSION*
```

In this stage we decide to implement our database on both local and GCP. We fixed the "Missed the Updates Relationship in the Schema." problem in the previous stage by specifying the Update/Delete conditions. The following code is the DDL we used.

```
1 CREATE TABLE USER(
2 user_id INT,
3 sex VARCHAR(255),
 4 country VARCHAR(255),
5 FirstName VARCHAR(255),
6 LastName VARCHAR(255),
 7 password VARCHAR(255),
8 age INT,
9 PRIMARY KEY(user_id)
10 );
11
12 CREATE TABLE CONDITIONS(
13 trackable_id INT,
14 name VARCHAR(255),
15 description VARCHAR(255),
16 treatment_options VARCHAR(255),
17 amount_of_patient INT,
18 average_age INT,
19 PRIMARY KEY(trackable_id)
20 );
21
22 CREATE TABLE SYMPTOMS(
23 trackable_id INT,
24 amount_of_patient INT,
25 average_age INT,
26 name VARCHAR(255),
27 location VARCHAR(255),
28 description VARCHAR(255),
29 PRIMARY KEY(trackable_id)
30 );
31
```

```
32 CREATE TABLE REPORTING(
  33 report_id INT,
34 report_type WARCHAR(255),
  35 reported_symptom VARCHAR(255),
36 reported_description VARCHAR(255),
  37 PRIMARY KEY(report_id)
  38 ):
  48 CREATE TABLE WISHLIST(
41 wishlist_id INT,
  42 illness_name VARCHAR(255),
43 illness_description VARCHAR(255),
  44 user_id INT,
  45 subscription_encil WACHAR(255),
  46 PRIMARY KEY(WISHILSELID),
47 FOREIGN KEY(user_id) REFERENCES USER(user_id) ON DELETE CASCADE ON UPDATE CASCADE
  49
  58 CREATE TABLE relate_toC
     condition_id INT,
 55 symptom_id INT,
53 condition_name VARCHAR(255),
54 symptom_name VARCHAR(255),
55 relation_count INT,
  56 PRIMARY KEY(condition_id, symptom_id),
57 FOREIGN KEY(condition_id) REFERENCES COMDITIONS(trockable_id) ON DELETE CASCADE ON UPDATE CASCADE,
     FOREIGN KEY(symptom_id) REFERENCES SYMPTOMS(trockable_id) ON DELETE CASCADE ON UPDATE CASCADE
  59 ):
 61.
61
62 CREATE TABLE diagnosed_with(
63 condition_id INT,
64 user_id int,
65 PRIMARY KEY(condition_id, user_id),
66 FOREIGN KEY(condition_id) REFERENCES CONDITIONS(trackable_id) ON DELETE CASCADE ON UPDATE CASCADE,
67
    FOREIGN KEY(user_id) REFERENCES USER(user_id) ON DELETE CASCADE ON UPDATE CASCADE
68 );
69
70 CREATE TABLE suffer_from(
71 user_id INT,
    symptom_id INT,
73 PRIMARY KEY(user_id, symptom_id),
74 FOREIGN KEY(user_id) REFERENCES USER(user_id) ON DELETE CASCADE ON UPDATE CASCADE,
    FOREIGN KEY(symptom_id) REFERENCES SYMPTOMS(trackable_id) ON DELETE CASCADE ON UPDATE CASCADE
76 );
```

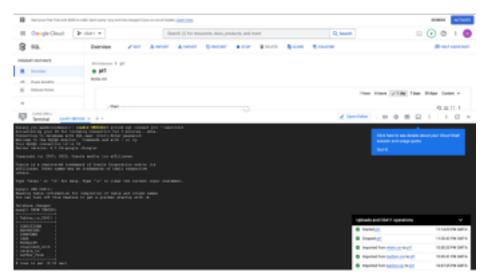
All of the tables in the database is shown here:

Here is some deep look into the structure of three of our tables:

```
amount_of_patient | int(11)
 average_age
                    varchar(255)
                  varchar(255)
 description
rows in set (0.00 sec)
 description
                   varchar(255)
 treatment_options | varchar(255)
 amount_of_patient | int(11)
 average_age
mysql> DESCRIBE relate_to;
Field
                    | Null | Key | Default | Extra |
 symptom_id | int(11) | NO
                           | PRI | NULL
 rows in set (0.00 sec)
mysql> DESCRIBE CONDITIONS;
 Field
                                    | Null | Key | Default | Extra |
                    | Type
 trackable_id
                                           | PRI | NULL
                      varchar(255)
 description
 treatment_options |
                      varchar(255)
 amount_of_patient |
                                                   NULL
 average_age
6 rows in set (0.04 sec)
mysql> DESCRIBE relate_to;
| Field
                                   | Null | Key | Default | Extra |
                  | Type
 condition_id
                                          PRI
  symptom_id
                    int(11)
                                            PRI
 condition_name | varchar(255)
 symptom_name
                  | varchar(255) |
 relation_count | int(11)
5 rows in set (0.00 sec)
```

mysql> DESCRIBE SYMPTOMS;

For GCP it's all the same, here's a screenshot of the console:



## 2. Data Pre-Processing

The dataset we used is rich in information yet lacks organization, so we spent a great amount of time pre-processing it to make it usable. We took the following steps to organize the dataset into separate clean data.

### **Conditions table and Symptoms table**

```
Total numer of unique conditions ('trackable_name') tracked are 9443
Total numer of unique symptoms ('trackable_name') tracked are 23157
```

Initially, we got 9443 unique conditions and 23157 symptoms in the dataset. To simplify the case, we choose the top 1000 conditions/symptoms by the number of patients reporting having those. Then we calculate the average age of those conditions/symptoms based on the original dataset.

Step 1: Cleaning up and splitting the dataset

```
import pandas as pd
import numpy as np
## STEP 1: CLEAN UP AND SPLIT THE DATASET

res = pd.read_csv('export.csv')

df = pd.DataFrame(res)

df['user_id'] = pd.Categorical(df['user_id'])

df['user_id'] = df.user_id.cat.codes

df["age"] = df.age.replace(0.0,np.nan)

# Save data with trackable_type=="Symptom" or trackable_type=="Condition"

df_filtered_Symptom = df[df.trackable_type=="Symptom"]

df_filtered_Symptom.to_csv('Symptom.csv')

df_filtered_Condition = df[df.trackable_type=="Condition"]

df_filtered_Condition.to_csv('Condition.csv')
```

Step 2: Fetch top 1000 symptoms (same for conditions)

```
# # STEP 2: GET TOP 1000 SYMPTOMS

18    res = pd_read_csv('Symptom.csv')

19    # Create a DataFrame and filter for rows with "trackable_type" equal to "Symptom"

20    df = pd_DataFrame(res[res.trackable_type=="Symptom"])

21    # Get the top 10 frequently occurring symptoms

22    output = df.trackable_name.value_counts().head(1000)

23    # Calculate the mean age for each symptom in the top 10 list

24    mean_age = df[df.trackable_name.isin(output.index)].groupby("trackable_name"].mean().sort_values(by="age")

25    # Print the mean age of each symptom in the top 1000 list, sorted by the top 1000 order

26    out = mean_age.loc[output.index]

27    out.to_csv('Sym1000age.csv')
```

Step 3: Count top 1000 conditions' number of patients (same for symptoms)

```
# STEP 4: GET TOP 1000 CONSTITUTE COUNTS

res = pd_read_cos("Condition.cos")

df = pd_tateFrame(res)
print("Total numer of unique Condition ("trackable_name") tracked are",df[df.trackable_type=="Condition"].trackable_name.numique())

count = df[df.trackable_type=="Condition"].trackable_name.salue_counts[].head[1000]

Step

Step
```

4: Combine the above data and round up data to make things neat. **Conditions-Symptoms** relation table

For simplicity, we randomly chose 400000 data from the original dataset. Then we generated a table for the correlated relationship with the Symptoms and Conditions by counting how many patients reported having both simultaneously. We chose the correlations that have more than 500 records.

```
res = [d_read_csv('export.csv']
df = pd_bateframe(res)
df = pd_bateframe(res)
df = pd_bateframe(res)
df = pd_bateframe(res)
df = ef_sample(s = sesses)
# Create two Dataframes, one for Conditions and one for Symptoms
conditions = df[dff['trackable_type'] == 'Condition'], renume[columns=('trackable_id': 'condition_id', 'trackable_name': 'condition_name'))
symptoms = df[dff['trackable_type'] == 'Symptom'], renume[columns=('trackable_id': 'symptom_id', 'trackable_name': 'symptom_name'))
# Norse the two Dataframes on the user_id column
merged = pd_merge(conditions, symptoms, se='aser_id')
# Group the merged Dataframe by user_id, condition_id, symptom_id, condition_name, and symptom_name, thes count the occurrences
counts = merged.groupsty[['user_id', 'condition_id', 'symptom_id', 'condition_name', 'symptom_name']).size().reset_indexiname='count')
# Filter the counts Dataframe to only include rows where count is greater than or equal to 2
correlated = counts[Counts['count'] >= 500]
print(correlated.shape[0])
correlated.ta_csv('relate.csv'']
```

### 3. Data Insertion

In this stage, we finished inserting data into table "CONDITIONS", "SYMPTOMS" nad "relate\_to". The following shows the SQL commands we used for our local database. For the GCP, we simply click import.

```
LOAD DATA LOCAL INFILE "/Users/mac/Desktop/topCon.csv" INTO TABLE CS411.CONDITIONS
FIELDS TERMINATED BY ','
LINES TERMINATED BY '\n'
GNORE 1 LINES;

LOAD DATA LOCAL INFILE "/Users/mac/Desktop/topSym.csv" INTO TABLE CS411.SYMPTOMS
FIELDS TERMINATED BY ','
LINES TERMINATED BY '\n'
GNORE 1 LINES;

LOAD DATA LOCAL INFILE "/Users/mac/Desktop/relate.csv" INTO TABLE CS411.relate_to
FIELDS TERMINATED BY ','
LINES TERMINATED BY ','
LINES TERMINATED BY ','
LINES TERMINATED BY '\n'
```

### Here are some screenshots of our database:

Insert 1000 rows into TABLE "CONDITIONS"

Insert 1000 rows into TABLE "SYMPTOMS"

Insert 1000+ rows into TABLE "relate\_to"

COUNT(*)   									
					sql> SELECT *	FROM relate_	to ORDER BY RAND() LIMIT 20;		
condition_id	symptom_id	condition_name		relation_count					
1199	242	Acid Reflux	Fatigue	5250					
1186	279	Mast Cell Activation Syndrome	Muscle spasms	572					
1174	121	joint pain	Joint pain	4389					
1132	331	POTS	muscle weakness	520					
421	13432	Gastroparesis	being sick	596					
80		Asthma	Constipation	638					
222	1890	Chronic hives (urticaria)	Nose sores	525					
		Depression	Chest pain	528					
	2762	Fibromyalgia	hand and wrist pain	504					
6154	12247	bowel not working very well	got shuff pouring out underneath	675					
532	54	Idiopathic hypersomnia	Depression	1089					
291	515	Depression	exhaustion	528					
1164	329	Fatigue	impaired cognition	1276					
80	7984	Asthma	GI Distress	596					
1201	674	Chronic Pain	Neuropathy	961					
	11505	Hepatitis C	How Do You Feel Today?	720					
1493	8518	ME/CFS	overexertion	1421					
269	243	Crohn's disease	Stomach Pain	532					
1158	142	OCD	Mood swings	912					
728	415	Osteoarthritis	Achiness	754					

### 4. Advanced Queries

# **QUERY1:** Condition and symptom names that are related to each other

This query finds the average age of patients who have at least two symptoms in common, along with the names of the conditions and symptoms. This is useful because we want to find patients that have similar histories and then we can extrapolate more information based off of those similar histories. Perhaps even find a cause for the symptoms or derive risk factors from them.

### The commands:

SELECT c.name AS condition\_name, s.name AS symptom\_name, AVG(s.average\_age) AS avg\_age

FROM CONDITIONS c JOIN relate to r1 ON c.trackable id = r1.condition id

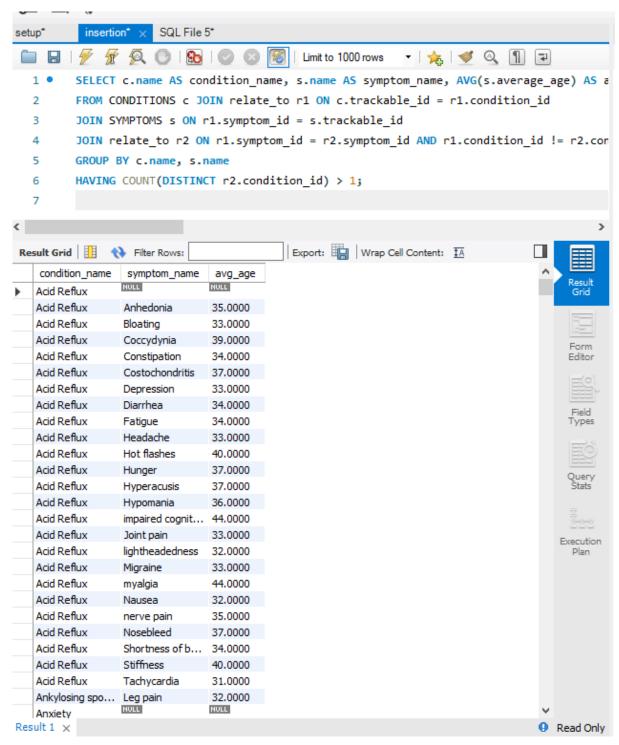
JOIN SYMPTOMS s ON r1.symptom id = s.trackable id

JOIN relate\_to r2 ON r1.symptom\_id = r2.symptom\_id AND r1.condition\_id != r2.condition\_id

GROUP BY c.name, s.name

HAVING COUNT(DISTINCT r2.condition\_id) > 1;

#### The results screenshot:



## **Indexing Analysis**

Default query:

```
Filter: (count(distinct relate_to.condition_id) > 1) (actual time=21.133..26.635 rows=956 loops=1)
  -> Group aggrègate: count(distinct relate_to.condition_id), avg(symptoms.average_age) (actual
 time=21.132..26.566 rows=1039 loops=1)
      -> Sort: c.`name`, s.`name` (actual time=21.117..21.801 rows=11736 loops=1)
-> Stream results (cost=1972.17 rows=4780) (actual time=0.041..13.004 rows=11736 loops=1)
-> Nested loop inner join (cost=1972.17 rows=4780) (actual time=0.039..10.026 rows=11736
                     -> Nested loop inner join (cost=1141.35 rows=1193) (actual time=0.032..2.646 rows=1193
                     loops=1)
                           -> Nested loop inner join (cost=723.80 rows=1193) (actual time=0.028..1.872 rows=1193
                          loops=1)
                                -> Ćovering index scan on r1 using rs_idx (cost=127.30 rows=1193) (actual
                               time=0.020..0.311 rows=1193 loops=1)
                               -> Single-row index lookup on c using PRIMARY (trackable_id=r1.condition_id) (
                               cost=0.40 rows=1) (actual time=0.001..0.001 rows=1 loops=1193)
                          -> Single-row index lookup on s using PRIMARY (trackable_id=r1.symptom_id) (cost=0.25
                     rows=1) (actual time=0.001..0.001 rows=1 loops=1193)
-> Filter: (r1.condition_id <> r2.condition_id) (cost=0.25 rows=4) (actual time=0.002..0.006 rows=10 loops=1193)
                          -> Covering index lookup on r2 using rs_idx (symptom_id=r1.symptom_id) (cost=0.25
                          rows=4) (actual time=0.002..0.005 rows=11 loops=1193)
```

So this is the EXPLAIN ANALYZE result of the original query.

*Index 1:* CREATE INDEX symptomname idx on symptoms(name);

```
Filter: (count(distinct relate_to.condition_id) > 1) (actual time=22.808..28.489 rows=956 loops=1) -> Group aggregate: count(distinct relate_to.condition_id), avg(symptoms.average_age) (actual
                      416 rows=1039 loops=1)
 time=22.8
       --> Sort: c. name`, s.`name` (actual time=22.791..23.526 rows=11736 loops=1)
-> Stream results (cost=1972.17 rows=4780) (actual time=0.069..14.320 rows=11736 loops=1)
-> Nested loop inner join (cost=1972.17 rows=4780) (actual time=0.065..11.225 rows=11736
                       -> Nested loop inner join (cost=1141.35 rows=1193) (actual time=0.057..3.207 rows=1193
                            -> Nested loop inner join (cost=723.80 rows=1193) (actual time=0.053..2.089 rows=1193
                           loops=1)
                                 -> Covering index scan on r1 using rs_idx (cost=127.30 rows=1193) (actual
                                time=0.037..0.330 rows=1193 loops=1)
                                -> Single-row index lookup on c using PRIMARY (trackable_id=r1.condition_id) (
                           cost=0.40 rows=1) (actual time=0.001..0.001 rows=1 loops=1193)

-> Single-row index lookup on s using PRIMARY (trackable_id=r1.symptom_id) (cost=0.25
                           rows=1) (actual time=0.001..0.001 rows=1 loops=1193)
                      -> Filter: (r1.condition_id <> r2.condition_id) (cost=0.25 rows=4) (actual
                      time=0.002..0.006 rows=10 loops=1193)
                           -> Covering index lookup on r2 using rs_idx (symptom_id=r1.symptom_id) (cost=0.25
                           rows=4) (actual time=0.002..0.005 rows=11 loops=1193)
```

To begin we make a simple index on the symptom name to see if it makes any difference. You can see that the times actually increase, but the costs themselves are the exact same as before. So this is not actually improving the efficiency of the query but seems to be making the times worse anyway.

*Index 2:* CREATE INDEX cidx on relate to(condition id, symptom id)

Let's try making an index for the relate\_to table instead. We will also try using two columns this time, condition id and symptom id

So immediately there appears to be a boost in overall time (from 21.1 seconds to 20.4), and also we see the costs have lowered (from 1972 to 1966 for the top level operation. Some of the other operations have also decreased (1141 to 1135 for the nested inner loop join), 723 to 717, etc. lots of slight improvements). The structure of the query still hasn't changed though, there are still 12 operations total.

Index 3: CREATE INDEX cdidx on c(condition\_id, symptom\_id)

With this last attempt, the time actually goes up again, but the costs are significantly lower (by almost 200 points). You can see that the first "Stream results" operation goes from 1966 to 1787, which is better than the previous

two indexes and also better than the original query as well. Similarly, all the other operations have much lower costs (except for the last few which seem to be constant at a cost of 0.25). But overall if we had to choose one query it would be this one since it is the one showing the biggest improvements from the original.

### **QUERY 2:** Select symptom that occurs in the most conditions

This query retrieves the possible symptoms in the database, ordered by the frequency of their occurrence in conjunction with the conditions. Since the relate\_to table, brings the symptom and condition data together, we utilized it to find how many instances of the symptoms occur across the various conditions using COUNT.

```
1 SELECT r.symptom_id , COUNT(r.symptom_id) AS cnt
2 FROM relate_to r
3 GROUP BY r.symptom_id
4 ORDER BY COUNT(r.symptom_id) DESC;
```

```
CLOUD SHELL
                    (cs411-380104) X
      Terminal
symptom id | cnt |
        98 |
               39 I
        152 |
               38 |
        242 |
               34
        45 I
               25
        63 |
               23
         8 I
        121
               22
        145
               21
        275 |
               21
               19
        54 |
        12 |
               17
        35 I
               17
        56 I
        399 I
               14
        257 |
               13
        515 |
               13
        191 |
               12
        329 |
               11
      2605 |
               11
        84 I
               10
        265 | 10 |
        681 |
               10 |
```

## **Indexing Analysis**

Default first:

So about 1.2 seconds

*Index 1:* CREATE INDEX rs idx on relate to(symptom id);

Let's try indexing symptom id:

```
# EXPLAIN

-> Sort: cnt DESC (actual time=0.518..0.527) ows=268 loops=1)

-> Stream results (cost=240.60 rows=268) (actual time=0.106..0.448 rows=268 loops=1)

-> Group aggregate: count(r.symptom_id) (cost=240.60 rows=268) (actual time=0.103..0.418 rows=268 loops=1)

-> Covering index scan on r using rs_idx (cost=121.30 rows=1193) (actual time=0.097..0.349 rows=1193 loops=1)
```

Okay, so the time dropped about 0.7 seconds. Quite good. *Index 2:* CREATE INDEX rc\_idx on relate\_to(condition\_id);

Well, there's not much else to index since in the query above we only really checked one field. So we will just try other fields in the relate\_to table and see if changing them gives any difference.

```
# EXPLAIN

-> Sort: cnt DESC (actual time=0.518..0.527) rows=268 loops=1)

-> Stream results (cost=240.60 rows=268) (actual time=0.106..0.448 rows=268 loops=1)

-> Group aggregate: count(r.symptom_id) (cost=240.60 rows=268) (actual time=0.103..0.418 rows=268 loops=1)

-> Covering index scan on r using rs_idx (cost=121.30 rows=1193) (actual time=0.097..0.349 rows=1193 loops=1)
```

Strangely enough, it seems to still speed up the queries.

Index 3: CREATE INDEX rsn\_idx on relate\_to(symptom\_name);

```
# EXPLAIN
-> Sort: cnt DESC (actual time=0.629..0.636 rows=268 loops=1)
   -> Stream results (cost=240.60 rows=268) (actual time=0.055..0.537 rows=268 loops=1)
   -> Group aggregate: count(r.symptom_id) (cost=240.60 rows=268) (actual time=0.053..0.490 rows=268 loops=1)
   -> Covering index scan on r using rs_idx (cost=121.30 rows=1193) (actual time=0.046..0.389 rows=1193 loops=1)
```

And for the final one, it's a bit slower at 0.629 sec. The costs for all three are the same, so there's not much to comment on there. The query itself is perhaps a little too simple to benefit much from indexing. Or perhaps, we should try combinations of columns from different tables. But with the current strategies, we don't seem to get much improvement.