## **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 VARCHAR(255),
 reported_symptom VARCHAR(255),
reported_description VARCHAR(255),
  37
     PRIMARY KEY(report_id)
  38 );
  40 CREATE TABLE WISHLISTC
  41 wishlist_id INT,
 42 illness_name VARCHAR(255),
43 illness_description VARCHAR(255),
  44 user_id INT,
  45 subscription_email VARCHAR(255),
      PRIMARY KEY(wishits_id),
FOREIGN KEY(user_id) REFERENCES USER(user_id) ON DELETE CASCADE ON UPDATE CASCADE
 47
 49
  50 CREATE TABLE relate_to(
 51 condition_id INT,
52 symptom_id INT,
 53 condition_name VARCHAR(255),
54 symptom_name VARCHAR(255),
 54 symptom_name Vakchak(255),
55 relation_count INT,
56 PRIMARY KEY(condition_id, symptom_id),
57 FOREIGN KEY(condition_id) REFERENCES CONDITIONS(trackable_id) ON DELETE CASCADE ON UPDATE CASCADE,
58 FOREIGN KEY(symptom_id) REFERENCES SYMPTOMS(trackable_id) ON DELETE CASCADE ON UPDATE CASCADE
  59
     );
 60
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
69
70 CREATE TABLE suffer_from(
71 user_id INT,
72 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,
75 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:

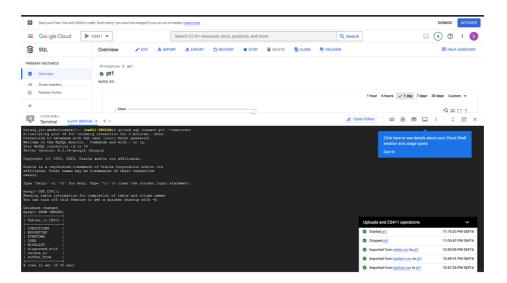
Here is some deep look into the structure of three of our tables:

```
mysql> DESCRIBE SYMPTOMS;
 Field
                                  | Null | Key | Default | Extra |
 trackable_id
                                   l no
                                         | PRI |
 amount_of_patient
 average_age
                     varchar(255)
 location
 description
6 rows in set (0.00 sec)
mysql> DESCRIBE ILLNESES;
 trackable_id
                                   NO
                                          | PRI |
                                                 NULL
                     varchar(255)
 description
 treatment_options | varchar(255)
 amount_of_patient
                                                 NULL
 average_age
6 rows in set (0.00 sec)
mysql> DESCRIBE relate_to;
 Field
            | Type
                      | Null | Key | Default | Extra |
 illness_id | int(11) | NO
                             | PRI
                                   | NULL
 symptom_id | int(11) | NO
                               PRI I
 rows in set (0.00 sec)
```

mysql> DESCRIBE COND	ITIONS;	<b>.</b>			
   Field		Null +	   Key	Default	   Extra   
trackable_id   name   description   treatment_options   amount_of_patient   average_age	int(11) varchar(255) varchar(255) varchar(255) int(11) int(11)	NO YES YES YES YES YES	PRI	NULL NULL NULL NULL NULL NULL	
6 rows in set (0.04 s	sec)				

```
mysql> DESCRIBE relate_to;
                                | Null | Key | Default | Extra |
| Field
                 | Type
| condition_id
                   int(11)
                                         PRI |
                                               NULL
| symptom_id
                   int(11)
                                  NO
                                         PRI
                                               NULL
                  varchar(255)
                                               NULL
condition_name |
| symptom_name
                  varchar(255)
                                  YES
                                               NULL
| relation_count | int(11)
                                  YES
                                               NULL
5 rows in set (0.00 sec)
```

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_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

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

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

df = pd.DataFrame(res[res.trackable_type=="Symptom"])

# Get the top 10 frequently occurring symptoms

utput = df.trackable_name.value_counts().head(1000)

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

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

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

out = mean_age.loc[output.index]

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

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

```
# STEP 4: GET TOP 1000 CONDITIONS COUNTS
res = pd.read_csv('Condition.csv')

df = pd.DataFrame(res)
print("Total numer of unique Condition ('trackable_name') tracked are",df[df.trackable_type=="Condition"].trackable_name.nunique())
count = df[df.trackable_type=="Condition"].trackable_name.value_counts().head(1000)
count.to_csv('topCon.csv')
```

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 = pd.read_csv('export.csv')
df = pd.DataFrame(res)
df = df.sample(n = 400000)
# Create two DataFrames, one for Conditions and one for Symptoms
conditions = df[df['trackable_type'] == 'Condition'].rename(columns={'trackable_id': 'condition_id', 'trackable_name': 'condition_name'})
symptoms = df[df['trackable_type'] == 'Symptom'].rename(columns={'trackable_id': 'symptom_id', 'trackable_name': 'symptom_name'})
# Merge the two DataFrames on the user_id column
merged = pd.merge(conditions, symptoms, on='user_id')
# Group the merged DataFrame by user_id, condition_id, symptom_id, condition_name, and symptom_name, then count the occurrences
counts = merged.groupby(['user_id', 'condition_id', 'symptom_id', 'condition_name', 'symptom_name']).size().reset_index(name='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.to_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'
IGNORE 1 LINES;

LOAD DATA LOCAL INFILE "/Users/mac/Desktop/relate.csv" INTO TABLE CS411.relate_to
FIELDS TERMINATED BY ','
LINES TERMINATED BY ','
INTO TABLE CS411.relate_to
FIELDS TERMINATED BY '\n'
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(*)				
1188				
row in set (0	.00 sec)			
		to ORDER BY RAND() LIMIT 20;		
		condition_name	+   symptom_name	relation_count
1199	242	Acid Reflux	   Fatigue	
1186	279	Mast Cell Activation Syndrome	Muscle spasms	
1174	121	joint pain	Joint pain	
1132	331	POTS	muscle weakness	
	13432	Gastroparesis	being sick	
	45	Asthma	Constipation	
	1890	Chronic hives (urticaria)	Nose sores	525
	35	Depression	Chest pain	
397	2762	Fibromyalgia	hand and wrist pain	504
6154	12247	bowel not working very well	got shuff pouring out underneath	
532	54	Idiopathic hypersomnia	Depression	1089
291	515	Depression	exhaustion	528
1164	329	Fatigue	impaired cognition	1276
80	7004	Asthma	GI Distress	506
1201	674	Chronic Pain	Neuropathy	961
485	11505	Hepatitis C	How Do You Feel Today?	720
1493	8518		overexertion	1421
	243	Crohn's disease	Stomach Pain	
1158	142	OCD	Mood swings	912
728	415	Osteoarthritis	Achiness	l 754 l

## 4. Advanced Queries

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

This query retrieves distinct pairs of condition and symptom names that are related to each other. It uses joins to combine the SYMPTOMS, relate\_to, and CONDITIONS tables, then groups the results by the name of the condition and symptom, and filters the results to only include rows where the trackable\_id of the condition equals the condition\_id of the relate\_to table. Finally, the results are ordered by the name of the condition.

The commands:

```
1 SELECT DISTINCT c.name, s.name
2 FROM SYMPTOMS s
3 JOIN relate_to r ON r.symptom_name = s.name
4 JOIN CONDITIONS c ON c.trackable_id = r.condition_id
5 GROUP BY c.name, s.name, c.trackable_id, r.condition_id
6 HAVING c.trackable_id = r.condition_id
7 ORDER BY c.name;
8
```

The results screenshot:

```
Chronic daily headaches
Chronic fatique syndrome
```

## **Indexing Analysis**

Default query:

```
# EXPLAIN
-> Sort: c.`name`, s.`name`, c.trackable_id, r.condition_id (actual time=6.750..6.802 rows=1057 loops=1)
-> Sort with duplicate removal: c.`name`, s.`name` (actual time=6.334..6.389 rows=1057 loops=1)
-> Table scan on <temporary> (cost=2.50..2.50 rows=0) (actual time=5.662..5.772 rows=1057 loops=1)
-> Temporary table with deduplication (cost=0.00..0.00 rows=0) (actual time=5.660..5.660 rows=1057 loops=1)
-> Filter: (c.trackable_id = r.condition_id) (actual time=3.842..4.912 rows=1057 loops=1)
-> Filter: (s.`name` = r.symptom_name) (cost=159924.51 rows=159459) (actual time=3.839..4.822 rows=1057 loops=1)
-> Inner hash join (<hash>(s.`name`)=<hash>(r.symptom_name)) (cost=159924.51 rows=159459) (actual time=3.837..4.578 rows=1057 loops=1)
-> Table scan on s (cost=0.03 rows=1593) (actual time=0.055..0.490 rows=1593 loops=1)
-> Hash
-> Nested loop inner join (cost=451.95 rows=1001) (actual time=0.091..3.393 rows=1193 loops=1)
-> Table scan on c (cost=101.60 rows=1001) (actual time=0.051..0.441 rows=1001 loops=1)
-> Index lookup on r using PRIMARY (condition_id=c.trackable_id) (cost=0.25 rows=1) (actual time=0.002..0.003 rows=1 loops=1001)
```

The costs have been highlighted along with the times and the number of rows/loops. Time is around 7 seconds. Now we will try three indexes.

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

```
# EXPLAIN

-> Sort: c. name', s. name', c.trackable_id, r.condition_id (actual time=7.070.7.122 rows=1057 loops=1)

-> Sort: c. name', s. name', c.trackable_id, r.condition_id (actual time=6.053..6.705 rows=1057 loops=1)

-> Table scan on <temporary> (cost=2.50..2.50 rows=0) (actual time=6.025..6.133 rows=1057 loops=1)

-> Temporary table with deduplication (cost=0.00..0.00 rows=0) (actual time=6.024..6.024 rows=1057 loops=1)

-> Filter: (c.trackable_id = r.condition_id) (actual time=0.085..5.313 rows=1057 loops=1)

-> Nested loop inner join (cost=1609.71 rows=2349) (actual time=0.084..5.228 rows=1057 loops=1)

-> Nested loop inner join (cost=538.85 rows=1193) (actual time=0.068..1.142 rows=1193 loops=1)

-> Filter: (r.symptom_name is not null) (cost=121.30 rows=1193) (actual time=0.046..0.466 rows=1193 loops=1)

-> Table scan on r (cost=121.30 rows=1193) (actual time=0.046..0.466 rows=1193 loops=1)

-> Single-row index lookup on c using PRIMARY (trackable_id=r.condition_id) (cost=0.25 rows=1) (actual time=0.000 rows=1 loops=1193)

-> Covering index lookup on s using symptomname_idx (name=r.symptom_name) (cost=0.70 rows=2) (actual time=0.003..0.003 rows=1 loops=1193)
```

With this first index we actually seemed to have made the performance worse (from 6.750..6.802 to 7.070..7.122). The inner hash join has been replaced with

two nested loop inner joins. We've also gotten rid of an extra table scan. However there are now **two** index row lookups required, which is maybe why it is taking longer.

*Index 2:* CREATE INDEX condition id idx on relate to(condition id);

Let's try making an index for the relate\_to table instead. First we should drop the previous index like so:

```
DROP INDEX symptomname_idx on symptoms;
```

```
# EXPLAIN

-> Sort: c.`name`, s.`name`, c.trackable_id, r.condition_id (actual time=3.837..3.889 rows=1057 loops=1)

-> Sort with duplicate removal: c.`name`, s.`name` (actual time=3.422..3.477 rows=1057 loops=1)

-> Table scan on <temporary> (cost=2.50..2.50 rows=0) (actual time=2.757..2.867 rows=1057 loops=1)

-> Temporary table with deduplication (cost=0.00..0.00 rows=0) (actual time=2.756..2.756 rows=1057 loops=1)

-> Filter: (c.trackable_id = r.condition_id) (actual time=1.204..2.141 rows=1057 loops=1)

-> Filter: (s.`name` = r.symptom_name) (cost=190599.27 rows=190045) (actual time=1.203..2.078 rows=1057 loops=1)

-> Inner hash join (<hash>(s.`name`)=<hash>(r.symptom_name)) (cost=190599.27 rows=190045) (actual time=1.202..1.882 rows=1057 loops=1)

-> Table scan on s (cost=0.03 rows=1593) (actual time=0.035..0.449 rows=1593 loops=1)

-> Hash

-> Nested loop inner join (cost=538.85 rows=1193) (actual time=0.065..0.876 rows=1193 loops=1)

-> Table scan on r (cost=121.30 rows=1193) (actual time=0.052..0.463 rows=1193 loops=1)

-> Single-row index lookup on c using PRIMARY (trackable_id=r.condition_id) (cost=0.25 rows=1) (actual time=0.000..0.000 rows=1 loops=1193)
```

So immediately there appears to be a boost in overall time, the 6-7 second times from before are around the 4 second range now. This query has the same structure as the default query except the last index lookup has been replaced with a single-row index lookup.

*Index 3:* CREATE INDEX conditions idx on conditions(name, trackable id);

Here we try doing two columns instead of 1, and we try it on the last table.

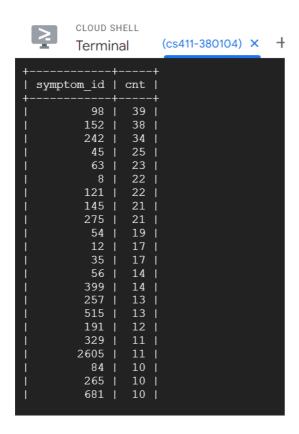
```
# EXPLAIN
-> Sort: c.`name`, s.`name`, c.trackable_id, r.condition_id (actual time_$.735..5.809) rows=1057 loops=1)
-> Sort with duplicate removal: c.`name`, s.`name` (actual time=5.126..5.214 rows=1057 loops=1)
-> Table scan on <temporary> (cost=2.50..2.50 rows=0) (actual time=4.044..4.216 rows=1057 loops=1)
-> Temporary table with deduplication (cost=0.00..0.00 rows=0) (actual time=4.042..4.042 rows=1057 loops=1)
-> Filter: (c.trackable_id = r.condition_id) (actual time=1.424..3.029 rows=1057 loops=1)
-> Filter: (s.`name` = r.symptom_name) (cost=190599.27 rows=190045) (actual time=1.423..2.915 rows=1057 loops=1)
-> Inner hash join (<hash>(s.`name`)=<hash>(r.symptom_name)) (cost=190599.27 rows=190045) (actual time=1.421..2.586 rows=1057 loops=1)
-> Table scan on s (cost=0.03 rows=1593) (actual time=0.035..0.763 rows=1593 loops=1)
-> Hash
-> Nested loop inner join (cost=538.85 rows=1193) (actual time=0.045..1.039 rows=1193 loops=1)
-> Table scan on r (cost=121.30 rows=1193) (actual time=0.036..0.504 rows=1193 loops=1)
-> Single-row index lookup on c using PRIMARY (trackable_id=r.condition_id) (cost=0.25 rows=1) (actual time=0.000..0.000 rows=1 loops=1193)
```

So, with this last attempt our times are just under 6 seconds. This is better than the default and attempt 1 but not as good as attempt 2. Also, since we used two columns this will be more work to update. The structure of the query is the same as attempt 2 and the default query, in that there the last operation is the only difference and it is now a single-row index lookup.

## **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;
```



## **Indexing Analysis**

Default first:

```
# EXPLAIN
-> Sort: cnt DESC (actual time=1.262..1.269 rows=268 loops=1)
-> Stream results (cost=240.60 rows=268) (actual time=0.400..0.754 rows=268 loops=1)
-> Group aggregate: count(r.symptom_id) (cost=240.60 rows=268) (actual time=0.395..0.721 rows=268 loops=1)
-> Covering index scan on r using symptom_id (cost=121.30 rows=1193) (actual time=0.260..0.526 rows=1193 loops=1)
```

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=8.518..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)
```

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 0.451..0.458 nows=268 loops=1)
-> Stream results (cost=240.60 rows=268) (actual time=0.047..0.390 rows=268 loops=1)
-> Group aggregate: count(r.symptom_id) (cost=240.60 rows=268) (actual time=0.044..0.361 rows=268 loops=1)
-> Covering index scan on r using rs_idx (cost=121.30 rows=1193) (actual time=0.039..0.294 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.