Data Analyst and Health Data Fellow Screening Exercise

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Exercise 1: An SQL query that will display the duplicate records.

Query:

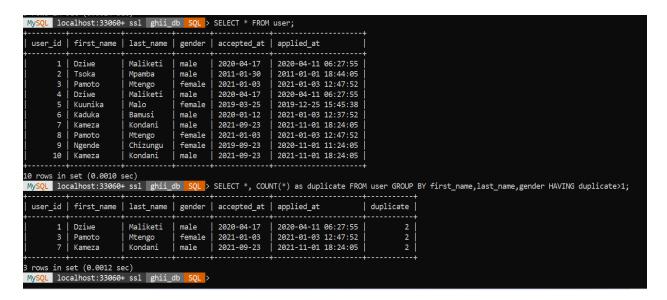
SELECT *, COUNT(*) as duplicate FROM user GROUP BY first_name, last_name, gender HAVING duplicate>1;

Explanation:

This SQL query selects all rows from the "user" table where there are duplicates in the combination of "first_name," "last_name," and "gender." It counts these duplicates and only returns rows where the count is greater than 1.

Solution:

The screenshot below shows the table before and after running the query;



Exercise 2:

(a) Two ways in which the query can be restructured (optimized) to run efficiently

1. Use of Common Table Expression (CTE) for the Subquery:

Instead of using a subquery within the main query's FROM clause, we can rewrite the subquery as a Common Table Expression (CTE). CTEs are often easier to read and maintain, and they sometimes result in better query optimization. Using a CTE can make the query more readable and potentially improve its performance.

2. Indexing:

Ensure that the relevant columns used for joining and filtering the data are properly indexed. In this case, we can add indexes to the following columns:

- Index the "person_id" column in the "Patient_Visit" table since it's used for grouping and joining.
- ➤ Index the "person_id" column in the "Diagnosis_Stage" table.
- Proper indexing can significantly improve the query's performance by reducing the time it takes to access and retrieve the required data.

(b) Query for the solution above:

Making the Subquery as Common Table Expression CTE:

```
WITH MaxVisitDates AS (

SELECT person_id, MAX(visit_date) AS visit_date

FROM Patient_Visit

GROUP BY person_id )

SELECT pd.person_id, pd.first_name, pd.last_name, mvd.visit_date

FROM Patient_Demographic pd

INNER JOIN MaxVisitDates mvd ON pd.person_id = mvd.person_id

WHERE pd.person_id IN (SELECT DISTINCT person_id FROM Diagnosis_Stage);
```

Exercise 3: Data Duplication

The below python script uses the panda library for reading data. The python script and the csv supporting docs can be found at https://github.com/JesseySitima/GHII-Data-analyst-Exercise.git

```
import pandas as pd
# task a: Read the CSV file
input_file = "client_purchases.csv"
df = pd.read_csv(input_file)
# task b: Find and identify duplicate records
duplicates = df[df.duplicated(subset=['First Name', 'Last Name', 'Email',
'Phone', 'Address', 'Zip Code', 'Gender', 'Date of Birth', 'Product',
'Quantity', 'Cost'])]
# task c: Remove the duplicate records
df = df.drop_duplicates(subset=['First Name', 'Last Name', 'Email',
'Phone', 'Address', 'Zip Code', 'Gender', 'Date of Birth', 'Product',
'Quantity', 'Cost'], keep='first')
# task d: Export the cleaned CSV file
cleaned_output_file = "client_purchases_deduplicated.csv"
df.to_csv(cleaned_output_file, index=False)
# task e: Identify unique clients and assign unique IDs
unique_clients = df[['First Name', 'Last Name', 'Email', 'Phone',
'Address', 'Zip Code', 'Gender', 'Date of
Birth']].drop_duplicates().reset_index(drop=True).reset_index()
unique_clients.rename(columns={'index': 'Client ID', 'First Name': 'First
Name', 'Last Name': 'Last Name', 'Email': 'Email', 'Phone': 'Phone',
'Address': 'Address', 'Zip Code': 'Zip Code', 'Gender': 'Gender', 'Date of
Birth': 'Date of Birth'}, inplace=True)
# Export the unique clients to a CSV file
unique_clients_output_file = "clients_unique.csv"
unique_clients.to_csv(unique_clients_output_file, index=False)
```

Exercise 4: Data Privacy and Security

- **a)** From the given deduplicated CVS, variables that qualify as Personal Identifiable Information are:
 - First Name
 - Last Name
 - > Email
 - Phone
 - Address
 - Zip Code
 - > Gender
 - Date of Birth
- **b)** Below is a python script using the Faker library which generates realistic but fake data.

```
import pandas as pd
from faker import Faker
# Initialize the Faker generator
fake = Faker()
# Read the CSV file with PII
input file = "client purchases deduplicated.csv"
df = pd.read_csv(input_file)
# Anonymize sensitive data
df['First Name'] = [fake.first_name() for _ in range(len(df))]
df['Last Name'] = [fake.last_name() for _ in range(len(df))]
df['Email'] = [fake.email() for _ in range(len(df))]
df['Phone'] = [fake.phone_number() for _ in range(len(df))]
df['Address'] = [fake.street_address() for _ in range(len(df))]
df['Zip Code'] = [fake.zipcode() for _ in range(len(df))]
df['Gender'] = [fake.random_element(elements=('Male', 'Female')) for
in range(len(df))]
df['Date of Birth'] = [fake.date_of_birth(minimum_age=18,
maximum_age=90).strftime('%m/%d/%Y') for _ in range(len(df))]
# Saving the anonymized data to a new CSV file
anonymized output file = "clients deidentified.csv"
df.to_csv(anonymized_output_file, index=False)
```